

Modern Control Theory

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Outline of Today's Lecture

- **Linearization**
- SS model of discrete control system
- Solution of Homogeneous State Equations



Linearization of a Nonlinear Function

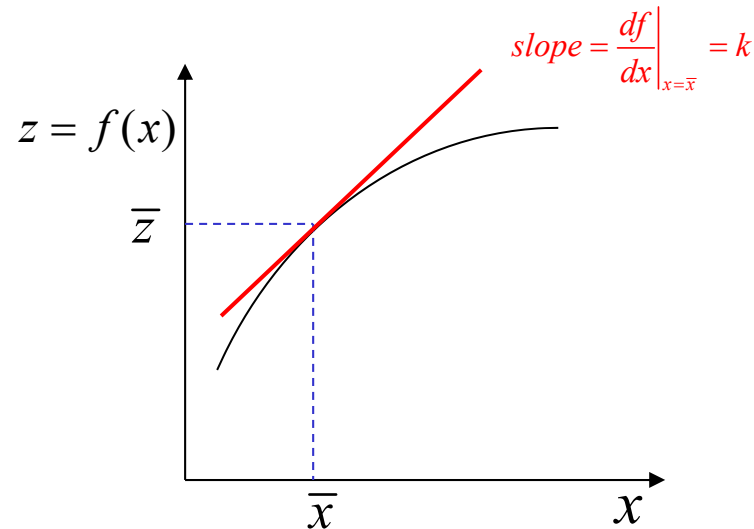
- Given a nonlinear function $z(t) = f(x(t))$, find a linear approximation.
- Basic idea is Taylor series expansion

$$z(t) = f(x(t)) = f(\bar{x}) + \left. \frac{df}{dx} \right|_{x=\bar{x}} (x - \bar{x}) + \left. \frac{d^2 f}{dx^2} \right|_{x=\bar{x}} \frac{(x - \bar{x})^2}{2!} + \dots$$
$$\approx f(\bar{x}) + \left. \frac{df}{dx} \right|_{x=\bar{x}} (x - \bar{x}) \equiv \bar{z} + k(x - \bar{x})$$



$$z - \bar{z} = k(x - \bar{x})$$

$$\Delta z = k \Delta x$$



Multiple Independent Variables

- Given a nonlinear function $z(t) = f(x(t), y(t))$, find a linear approximation.

$$z(t) = f(\bar{x}, \bar{y}) + \left. \frac{\partial f}{\partial x} \right|_{x=\bar{x}, y=\bar{y}} (x - \bar{x}) + \left. \frac{\partial f}{\partial y} \right|_{x=\bar{x}, y=\bar{y}} (y - \bar{y}) + \dots$$

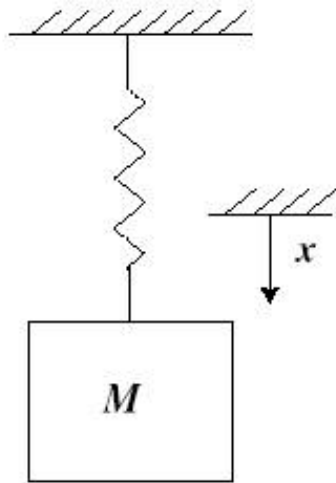
$$\approx f(\bar{x}, \bar{y}) + \left. \frac{\partial f}{\partial x} \right|_{x=\bar{x}, y=\bar{y}} (x - \bar{x}) + \left. \frac{\partial f}{\partial y} \right|_{x=\bar{x}, y=\bar{y}} (y - \bar{y}) \equiv \bar{z} + a(x - \bar{x}) + b(y - \bar{y})$$

$$\Delta z = a\Delta x + b\Delta y$$

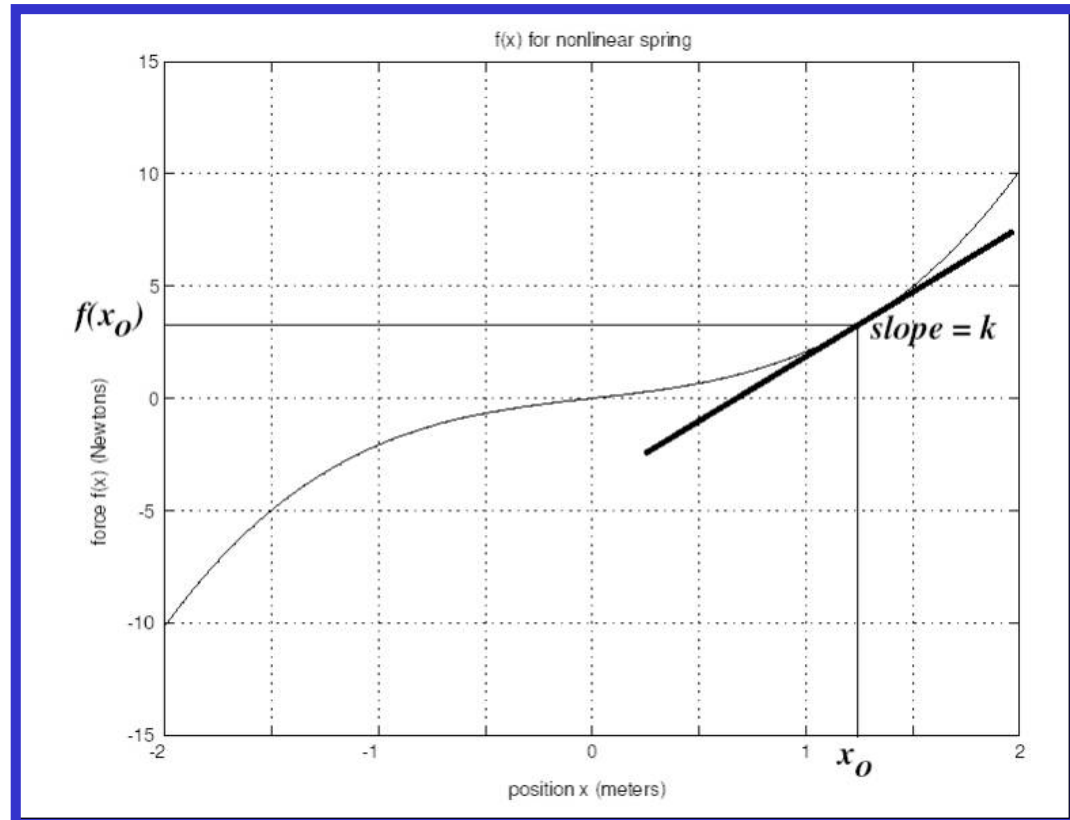


Ex.1 Nonlinear spring

1.1 Nonlinear spring



$$f(x) \approx f(x_o) + k\delta x$$



where k is the slope of the force/displacement curve at the operating point x_o .



Ex.1 Nonlinear spring (Cont)

The equilibrium is where the spring force balances gravity.

$$\dot{x}_o = 0$$

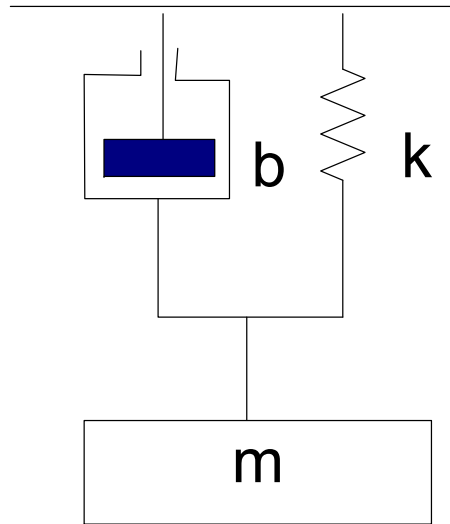
$$\ddot{x}_o = 0$$

$$Mg - f(x_o) = 0$$

$$f(x_o) = Mg$$



Ex.1 Nonlinear spring (Cont)



$$Mg - b\dot{x} - f(x) = M\ddot{x}$$

At the operating point x_o , the extension of the spring where $f(x_o) = Mg$. we do the Taylor expansion:

$$Mg - b(x_o + \delta x) - \left(f(x_o) + \frac{\partial f}{\partial x} \Big|_{x=x_o} \cdot \delta x \right) = M(x_o + \delta x)$$

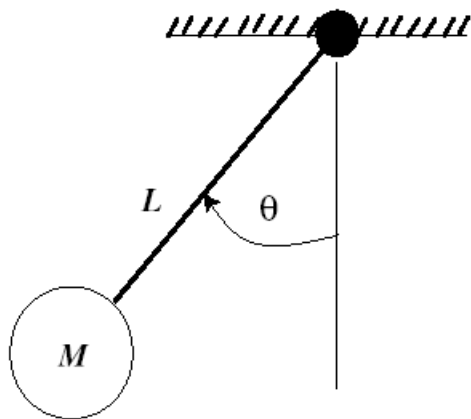
$$Mg - b\delta\dot{x} - f(x_o) - k\delta x = M\delta\ddot{x}$$

$$M\delta\ddot{x} + b\delta\dot{x} + k\delta x = 0$$



Ex.2 Pendulum

The torque is a nonlinear function of the angle .
We can use the operating point $\theta_0 = 0$, where the pendulum is hanging straight down.



$$f(\theta) = \tau = MgL \sin \theta$$

$$\tau \approx MgL \sin \theta_o + \left. \frac{\partial}{\partial \theta} (MgL \sin \theta) \right|_{\theta=\theta_o} \cdot \delta \theta$$

$$\approx 0 + MgL \cos \theta_o \delta \theta$$

$$\approx MgL \delta \theta$$



Ex.2 Pendulum (Cont.)

The equilibrium point:

$$\Sigma \text{torques} = J\ddot{\theta} = \tau - MLg \sin \theta$$

$$\dot{\theta}_o = 0$$

$$\ddot{\theta}_o = 0$$

$$\tau_o - MgL \sin \theta_o = 0$$

If there is no nominal applied torque, $\tau_o = 0$, then we have

$$\theta_o = 0 \quad \mathbf{OR} \quad \theta_o = \pi$$



Ex.2 Pendulum (Cont.)

$$\tau - MgL \sin \theta = ML^2 \ddot{\theta}$$

$$\theta \rightarrow \theta_o + \delta\theta, \tau \rightarrow \tau_o + \delta\tau$$

$$(\tau_o + \delta\tau) - (MgL \sin \theta_o + MgL \cos \theta_o \cdot \delta\theta) = ML^2 \ddot{\theta}_o + \delta\theta$$

$$\delta\tau - MgL \cos \theta_o \cdot \delta\theta = ML^2 \ddot{\delta\theta}$$

(1) Hanging pendulum: $\theta_o = 0$

$$ML^2 \ddot{\delta\theta} + MgL \delta\theta = \delta\tau$$

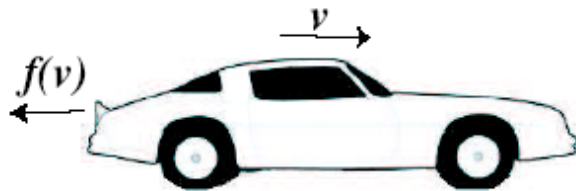
(2) Inverted pendulum: $\theta_o = \Pi$

$$ML^2 \ddot{\delta\theta} - MgL \delta\theta = \delta\tau$$



Ex.3 Aerodynamic drag

The force f is a nonlinear function of the velocity v ; typically, f is proportional to v^2 . We use the operating point v_o .



$$f(v) = av^2$$

$$f(v) \approx f(v_o) + \left. \frac{\partial f}{\partial v} \right|_{v=v_o} \cdot \delta v$$

$$= av_o^2 + 2av|_{v=v_o} \cdot \delta v$$

$$= av_o^2 + (2av_o)\delta v$$



Ex.3 Aerodynamic drag (Cont.)

The equilibrium point:

$$\Sigma \text{forces} = M\dot{v} = F - f(v)$$

For $v_o = \text{constant}$ to be a solution, we have

$$\dot{v}_o = 0$$

$$F_o - f(v_o) = 0$$

$$F_o - av_o^2 = 0$$

$$v_o = \sqrt{\frac{F_o}{a}}$$



Ex.3 Aerodynamic drag (Cont.)

$$F - av^2 = M\dot{v}$$

$$v \rightarrow v_o + \delta v, F \rightarrow F_o + \delta F$$

$$(F_o + \delta F) - (av_o^2 + (2av_o)\delta v) = M(v_o + \delta v)$$

$$F_o + \delta F - av_o^2 - (2av_o)\delta v = M\delta\dot{v}$$

Now, noting that $F_o = av_o^2$, we get

$$M\delta\dot{v} + (2av_o)\delta v = \delta F$$

Note that $\delta v = 0, \delta F = 0$ is a solution of the linearized equation.



Equilibrium Point

- In the application to dynamic systems, linearizations were usually performed at an “equilibrium point” for a dynamic system, i.e., for a nonlinear system, $\dot{X} = F(X, U)$ at a point (X_0, U_0) such that $F(X_0, U_0) = 0$



$$\dot{X} = 0$$



Linearization in state space model

-The time-varying system and Nonlinear system

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

$$y = c(t)x + D(t)u$$

$$\dot{x} = f(x, u, t)$$

$$y = g(x, u, t)$$

$$\dot{x}_0 = f(x_0, u_0) \quad \text{Taylor series expansion at equilibrium point}$$

$$f(x, u) = f(x_0, u_0) + \left. \frac{\partial f}{\partial x} \right|_{x_0, u_0} \delta x + \left. \frac{\partial f}{\partial u} \right|_{x_0, u_0} \delta u + \alpha(\delta x, \delta u)$$

$$g(x, u) = g(x_0, u_0) + \left. \frac{\partial g}{\partial x} \right|_{x_0, u_0} \delta x + \left. \frac{\partial g}{\partial u} \right|_{x_0, u_0} \delta u + \beta(\delta x, \delta u)$$



Linearization in state space model

$$\frac{\partial f}{\partial x} = \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} & \dots & \frac{\partial f_1}{\partial x_n} \\ \frac{\partial f_2}{\partial x_1} & \frac{\partial f_2}{\partial x_2} & \dots & \frac{\partial f_2}{\partial x_n} \\ \dots & \dots & \dots & \dots \\ \frac{\partial f_n}{\partial x_1} & \frac{\partial f_n}{\partial x_2} & \dots & \frac{\partial f_n}{\partial x_n} \end{bmatrix}$$

$$\frac{\partial g}{\partial x} = \begin{bmatrix} \frac{\partial g_1}{\partial x_1} & \frac{\partial g_1}{\partial x_2} & \dots & \frac{\partial g_1}{\partial x_n} \\ \frac{\partial g_2}{\partial x_1} & \frac{\partial g_2}{\partial x_2} & \dots & \frac{\partial g_2}{\partial x_n} \\ \dots & \dots & \dots & \dots \\ \frac{\partial g_m}{\partial x_1} & \frac{\partial g_m}{\partial x_2} & \dots & \frac{\partial g_m}{\partial x_n} \end{bmatrix}$$

$$\frac{\partial f}{\partial u} = \begin{bmatrix} \frac{\partial f_1}{\partial u_1} & \frac{\partial f_1}{\partial u_2} & \dots & \frac{\partial f_1}{\partial u_r} \\ \frac{\partial f_2}{\partial u_1} & \frac{\partial f_2}{\partial u_2} & \dots & \frac{\partial f_2}{\partial u_r} \\ \dots & \dots & \dots & \dots \\ \frac{\partial f_n}{\partial u_1} & \frac{\partial f_n}{\partial u_2} & \dots & \frac{\partial f_n}{\partial u_n} \end{bmatrix}$$

$$\frac{\partial g}{\partial u} = \begin{bmatrix} \frac{\partial g_1}{\partial u_1} & \frac{\partial g_1}{\partial u_2} & \dots & \frac{\partial g_1}{\partial u_r} \\ \frac{\partial g_2}{\partial u_1} & \frac{\partial g_2}{\partial u_2} & \dots & \frac{\partial g_2}{\partial u_r} \\ \dots & \dots & \dots & \dots \\ \frac{\partial g_m}{\partial u_1} & \frac{\partial g_m}{\partial u_2} & \dots & \frac{\partial g_m}{\partial u_r} \end{bmatrix}$$



Linearization in state space model

$$\delta \dot{x} = \dot{x} - \dot{x}_0 = \left. \frac{\partial f}{\partial x} \right|_{x_0, u_0} \delta x + \left. \frac{\partial f}{\partial u} \right|_{x_0, u_0} \delta u$$

$$\delta y = y - y_0 = \left. \frac{\partial g}{\partial x} \right|_{x_0, u_0} \delta x + \left. \frac{\partial g}{\partial u} \right|_{x_0, u_0} \delta u$$

$$\hat{x} = \delta x \quad \hat{u} = \delta u \quad \hat{y} = \delta y$$

$$\left. \frac{\partial f}{\partial x} \right|_{x_0, u_0} = A \quad \left. \frac{\partial f}{\partial u} \right|_{x_0, u_0} = B$$

$$\left. \frac{\partial g}{\partial x} \right|_{x_0, u_0} = C \quad \left. \frac{\partial g}{\partial u} \right|_{x_0, u_0} = D$$

$$\dot{\hat{x}} = A\hat{x} + B\hat{u}$$

$$\hat{y} = C\hat{x} + D\hat{u}$$



Example

- Linearization at $x_0=0$

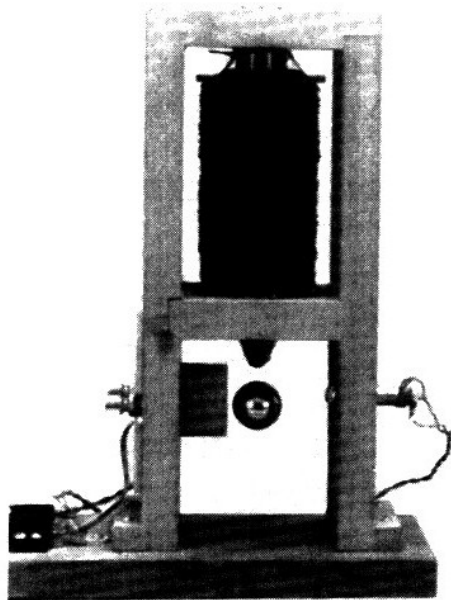
$$\dot{x}_1 = x_2$$

$$\dot{x}_2 = x_1 + x_2 + x_2^3 + 2u$$

$$y = x_1 + x_2^2$$



Ex. 2-22 (Franklin. pp. 70)



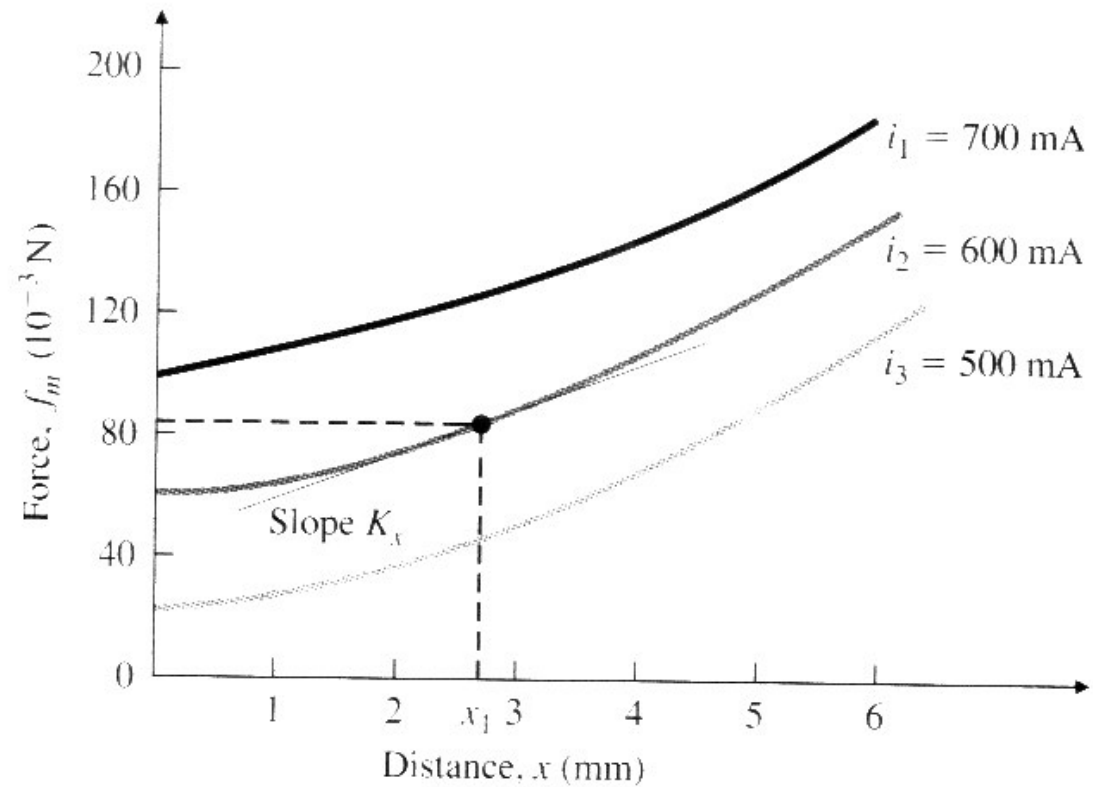
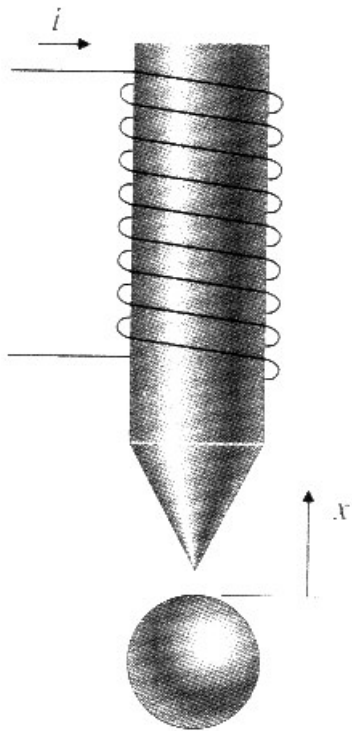
The simplified version of a magnetic bearing that can be built in a laboratory is shown in figure, where one electromagnet is used to levitate a metal ball. The equation of motion of the ball is:

$$m\ddot{x} = f_m(x, i) - mg$$

Where the force f_m is caused by the field of the electromagnet falls off with an inverse square relationship to the distance from the magnet. A experimental curves for a ball is shown in figure.



Ex. 2-22 (Cont.)



Ex. 2-22 (Cont.)

At the value for the current of $i_2 = 600\text{mA}$ and magnetic force f_m just cancels the gravity force $mg = 82 \times 10^{-3}\text{N}$. Therefore the point (x_1, i_2) represents an equilibrium. Using the data, find the linearized equations of motion about the equilibrium point.

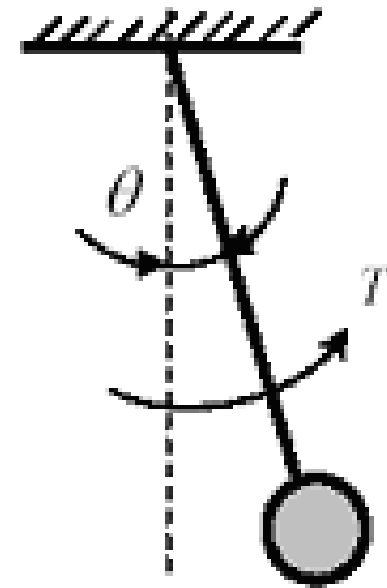


Keynotes-1: Linearization

Observe the pendulum shown in figure, moving in a gaseous fluid in which the aerodynamic drag is proportional to the square of the velocity of the body that moves through it, with b being the constant of proportionality i.e. aerodynamic drag is equal to

$$F_d = b(l\dot{\theta})^2 = bl^2\dot{\theta}^2$$

where θ is the angular pendulum distance away from its lowest point and l is the length of the pendulum). The pendulum is assumed to be moving only under the influence of gravity, with g being the gravity constant (around 9.8 N/Kg).



Keynotes-1: Linearization

- 1. Write out equation of motion in terms of the angle θ , denoting angular pendulum distance away from its lowest point. (8 points)**
- 2. Identify the equilibrium point. (6 points)**
- 3. Linearize equations of motion obtained in part (1) around the angle $\theta_0=0$, increment of the torque T as the system input, and increment of the angle θ as the system output. (10 points)**

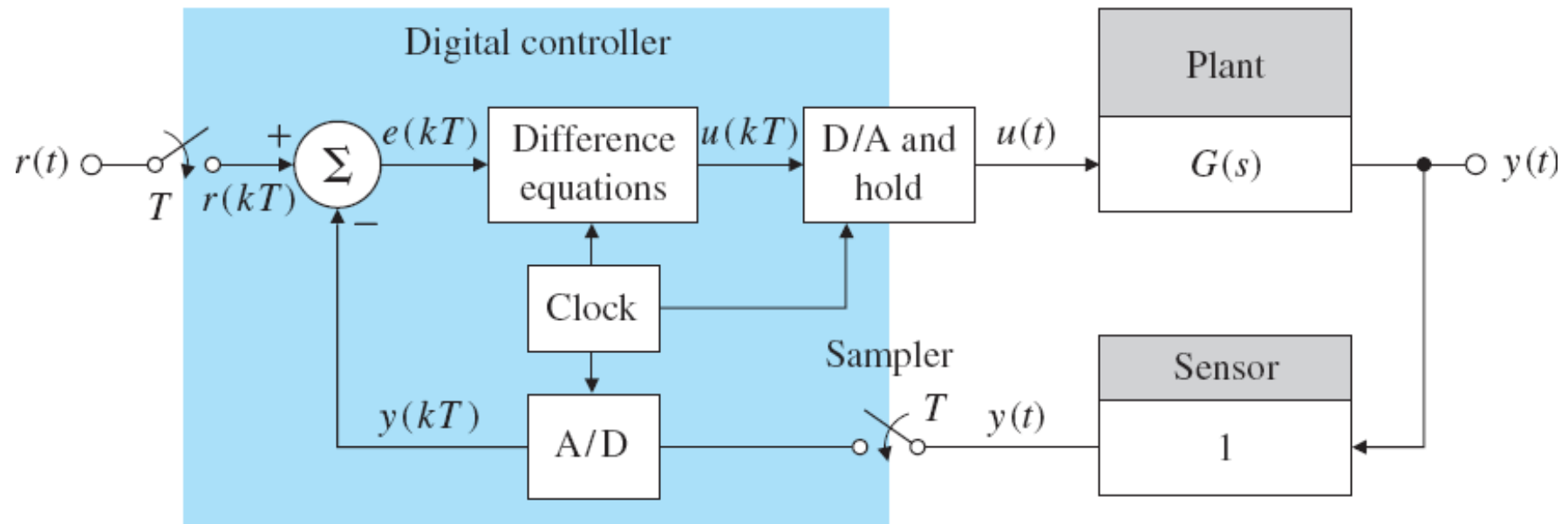
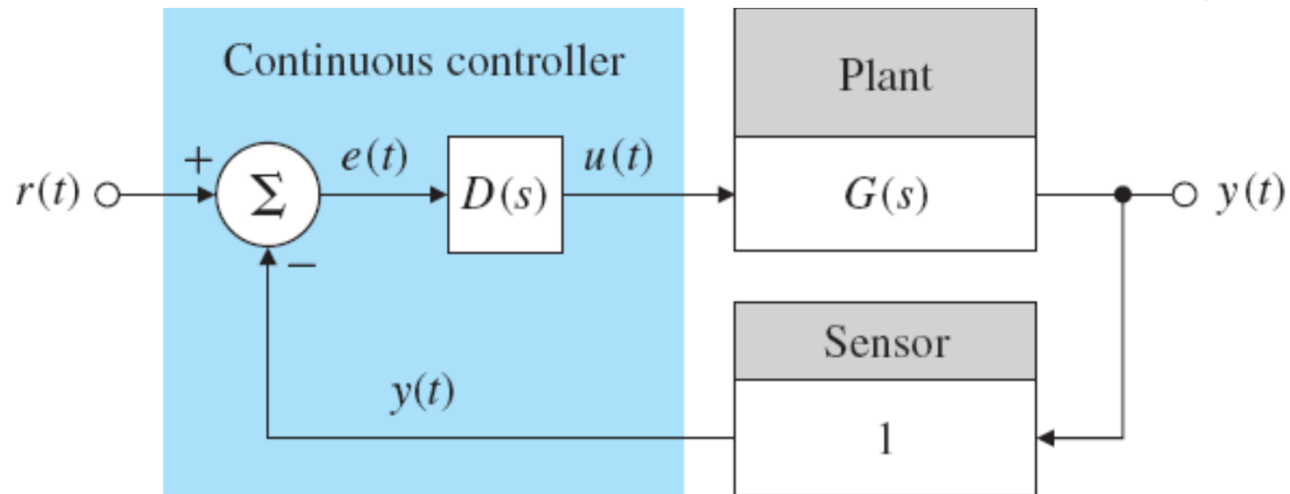


Outline of Today's Lecture

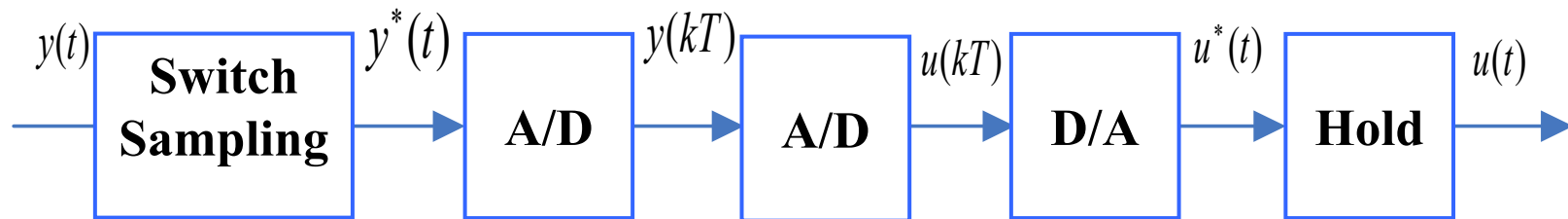
- Linearization
- **SS model of discrete control system**
- Solution of Homogeneous State Equations



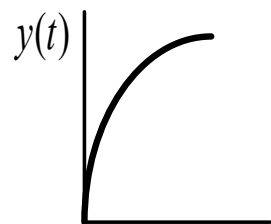
Brief Introduction of Discrete Control System



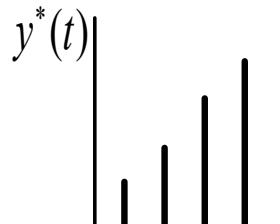
Introduction on Digital control



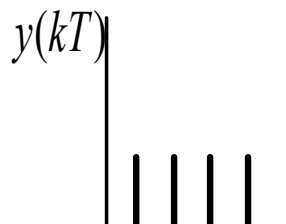
Zero-order hold (ZOH)



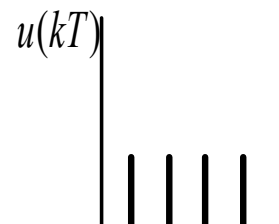
Continuous-time signal



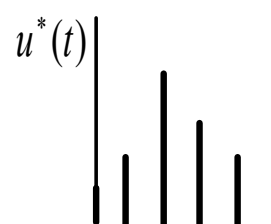
Sampled-data signal



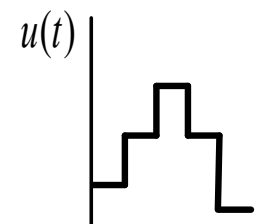
Digital signal



Digital signal



Digital signal



Digital signal



- **Sample period**
- **Sample rate**

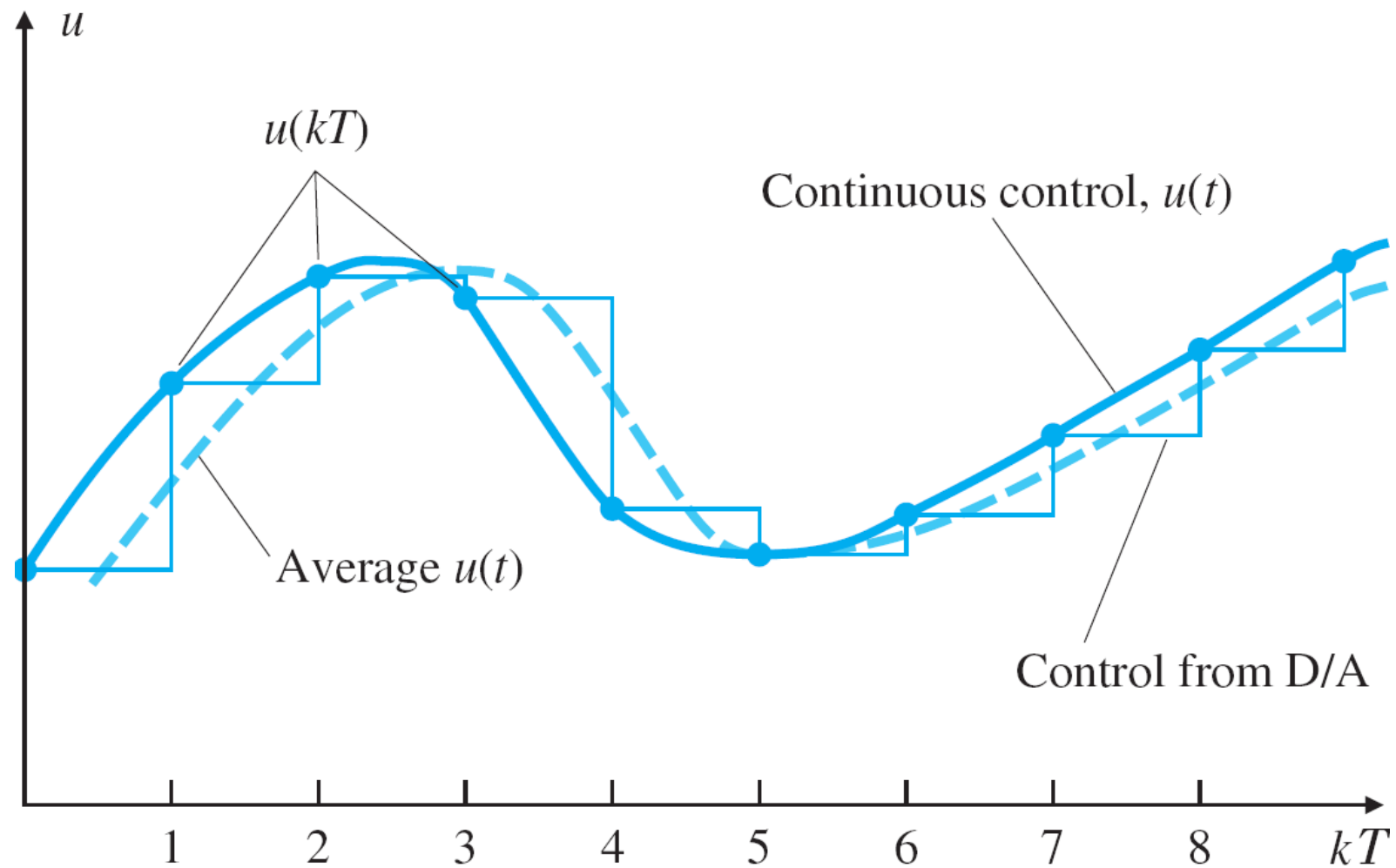
Brief Introduction of Discrete Control System

- Digital Control Systems operate on **samples** of the sensed plant output **rather than** on the continuous signal.
- Digital Control Systems are represented by **algebraic recursive equations** called **difference equations**.

$$\begin{aligned} & y(k+n) + a_{n-1}y(k+n-1) + \dots + a_1y(k+1) + a_0y(k) \\ & = b_nu(k+n) + b_{n-1}u(k+n-1) + \dots + b_1u(k+1) + b_0u(k) \end{aligned}$$



Brief Introduction of Discrete Control System



Brief Introduction of Discrete Control System

- Continuous-time signal
- Discrete-time signal
- Sampled-data signal
- Digital signal
- Sample period
- Sample rate
- Zero-order hold (ZOH)
- Discrete control system



State Space Representation for Discrete Control System

传递函数为:

$$W(z) = \frac{Y(z)}{U(z)} = \frac{b_n z^n + b_{n-1} z^{n-1} + \dots + b_1 z + b_0}{z^n + a_{n-1} z^{n-1} + \dots + a_1 z + a_0}$$

等效的状态空间表达式:

$$x(k+1) = Gx(k) + hu(k)$$

$$y(k) = C^T x(k) + du(k)$$



State Space Representation for

Discrete Control System

根据连续系统的推导，从模拟结构图可推导一种实现。

$$x(k+1) = \begin{bmatrix} 0 & 1 & 0 & \cdots & 0 \\ 0 & 0 & 1 & \cdots & 0 \\ \vdots & \vdots & \vdots & & \vdots \\ 0 & 0 & 0 & \cdots & 1 \\ -a_0 & -a_1 & -a_2 & \cdots & -a_{n-1} \end{bmatrix} x(k) + \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \\ 1 \end{bmatrix} u(k)$$

$$y(k) = [b_0 - a_0 b_n \quad b_1 - a_1 b_n \quad \cdots \quad b_{n-1} - a_{n-1} b_n] x(k)$$



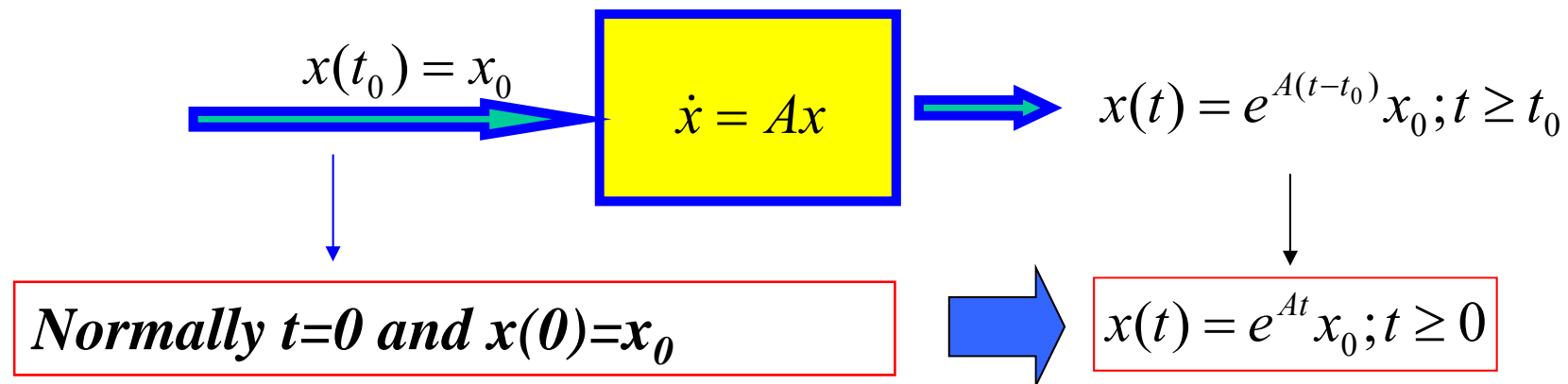
Outline of Today's Lecture

- Linearization
- SS model of discrete control system
- **Solution of Homogeneous State Equations**



Solution of Homogeneous State Equations

- The evolution of state variable under the initial state by assuming no input.



Review: the solution of scalar differential equation:

$$y^{(n)} + a_{n-1}y^{(n-1)} + a_{n-2}y^{(n-2)} + \dots + a_0y = b_m u^m + b_{m-1}u^{m-1} + \dots + b_0u$$



Solution of Homogeneous State Equations

$$y^{(n)} + a_{n-1}y^{(n-1)} + a_{n-2}y^{(n-2)} + \dots + a_0y = b_mu^m + b_{m-1}u^{m-1} + \dots + b_0u$$

homogeneous solution

$$y^{(n)} + a_{n-1}y^{(n-1)} + a_{n-2}y^{(n-2)} + \dots + a_0y = 0$$

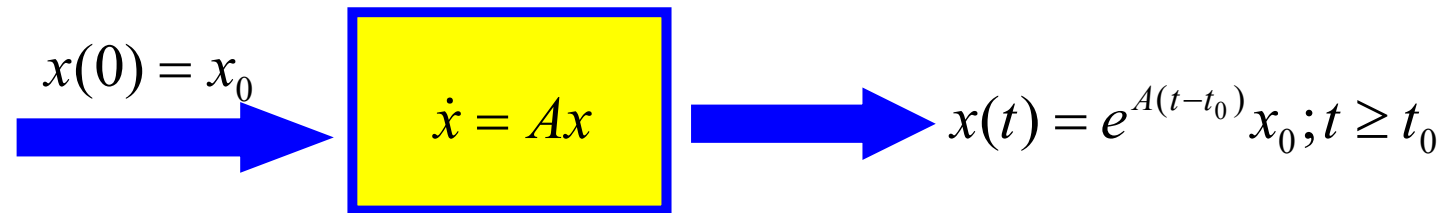
particular solution

$$y^{(n)} + a_{n-1}y^{(n-1)} + a_{n-2}y^{(n-2)} + \dots + a_0y = b_mu^m + b_{m-1}u^{m-1} + \dots + b_0u$$

*Solution of scalar differential equation
= homogeneous solution + one particular solution*



Solution of Homogeneous State Equations

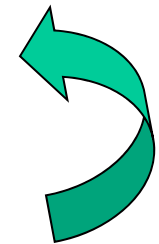


Firstly we consider
the scalar
differential
equation :
Suppose the solution has
a form of a polynomial of
t

$$\dot{x} = ax$$

$$x(t) = b_0 + b_1t + \dots + b_k t^k + \dots$$

$$\dot{x}(t) = b_1 + 2b_2t \dots + kb_k t^{k-1} + \dots$$

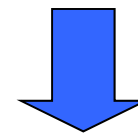


Solution of Homogeneous State Equations

Compare the coefficients of equal power of t :

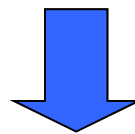
$$\left\{ \begin{array}{l} b_1 = ab_0 \\ b_2 = \frac{1}{2}ab_1 = \frac{1}{2}a^2b_0 \\ b_3 = \frac{1}{3}ab_2 = \frac{1}{3 \times 2}a^3b_0 \\ \dots \\ b_k = \frac{1}{k}ab_{k-1} = \frac{1}{k!}a^kb_0 \\ \dots \end{array} \right.$$

$$x(t) = b_0 + b_1t + \dots + b_kt^k + \dots$$



$$t=0, x(t)=x(0)$$

$$b_0 = x(0)$$



$$x(t) = (1 + at + \frac{1}{2!}a^2t^2 + \dots + \frac{1}{k!}a^kt^k + \dots)x(0)$$

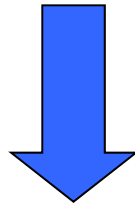


Important Properties of State Transition Matrix

For equation of Matix A $\dot{x} = Ax$

$$x(t) = (I + At + \frac{1}{2!} A^2 t^2 + \dots + \frac{1}{k!} A^k t^k + \dots)x(0)$$

**N × N
Matrix**



Define Matrix Exponential

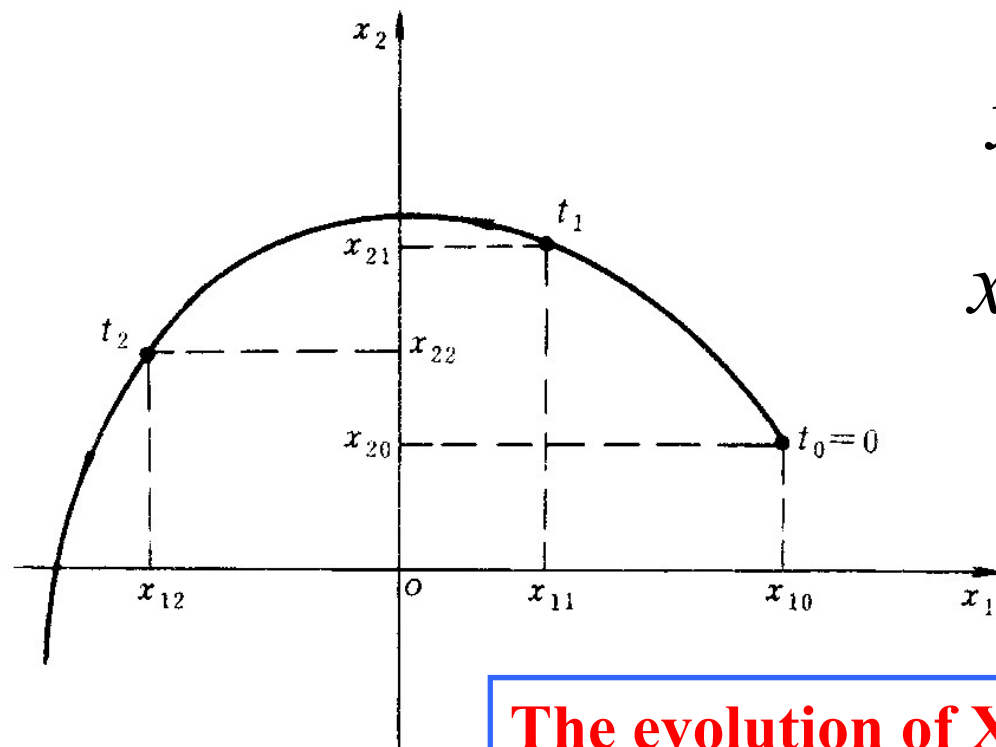
$$e^{At} = I + At + \frac{1}{2!} A^2 t^2 + \dots + \frac{1}{k!} A^k t^k + \dots$$

State Transition Matrix



Important Properties of State Transition Matrix

$$\dot{x} = Ax \xrightarrow{\text{define}} e^{At} \rightarrow \Phi(t)$$



$$x(t) = \Phi(t)x_0$$

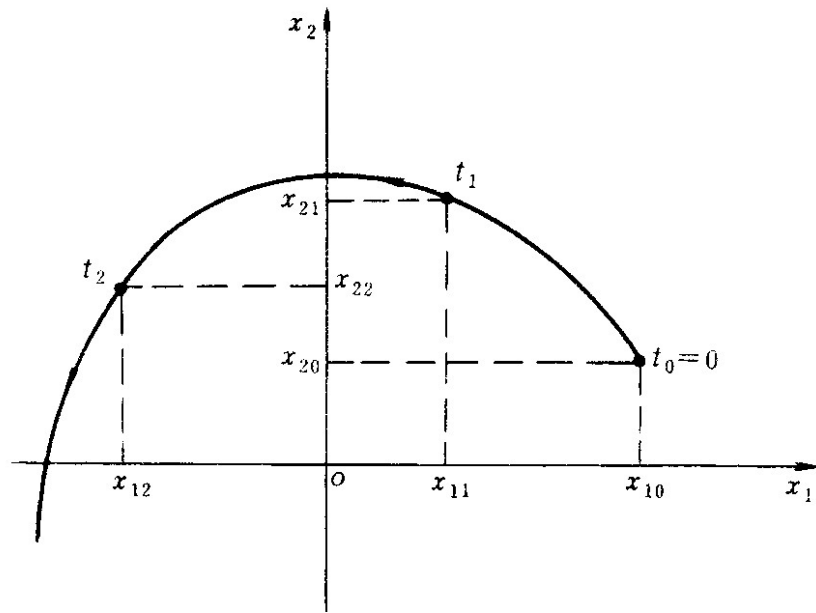
$$x(t_2) = \Phi(t_2 - t_1)x(t_1)$$



The evolution of $X(t)$ can be solved by the state transition matrix and any given state.



Important Properties of State Transition Matrices



$$1. \Phi(t + \tau) = \Phi(t)\Phi(\tau)$$

$$2. \Phi(0) = I$$

$$3. [\Phi(t)]^{-1} = \Phi(-t)$$

$$4. [\Phi(t)]^n = \Phi(nt)$$

$$5. \Phi(t_2 - t_1)\Phi(t_1 - t_0) = \Phi(t_2 - t_0) \\ = \Phi(t_1 - t_0)\Phi(t_2 - t_1)$$



Laplace transform Approach to the solution of Homogeneous Equations

$$\dot{x} = Ax$$



$$sX(s) - x(0) = AX(s)$$



$$(sI - A)X(s) = x(0)$$



$$X(s) = (sI - A)^{-1} x(0)$$



$$X(s) = L^{-1}[(sI - A)^{-1}]x(0)$$

$$\Phi(t) = L^{-1}[(sI - A)^{-1}]$$



Laplace transform Approach to the solution of Homogeneous Equations

Example: Obtain the state transition matrix $\phi(t)$ of the following system:

$$A = \begin{bmatrix} 0 & 1 \\ -2 & -3 \end{bmatrix}$$

Solution: The state transition Matrix $\phi(t)$ is given by:

$$\Phi(t) = L^{-1}[(sI - A)^{-1}]$$

$$sI - A = \begin{bmatrix} s & 0 \\ 0 & s \end{bmatrix} - \begin{bmatrix} 0 & 1 \\ -2 & -3 \end{bmatrix} = \begin{bmatrix} s & -1 \\ 2 & s+3 \end{bmatrix}$$



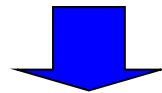
Laplace transform Approach to the solution of Homogeneous Equations

$$\begin{aligned}(sI - A)^{-1} &= \begin{bmatrix} s & -1 \\ 2 & s+3 \end{bmatrix}^{-1} = \frac{1}{(s+1)(s+2)} \begin{bmatrix} s+3 & 1 \\ -2 & s \end{bmatrix} \\&= \begin{bmatrix} \frac{s+3}{(s+1)(s+2)} & \frac{1}{(s+1)(s+2)} \\ \frac{-2}{(s+1)(s+2)} & \frac{s}{(s+1)(s+2)} \end{bmatrix} \\&= \begin{bmatrix} \frac{2}{(s+1)} - \frac{1}{(s+2)} & \frac{1}{(s+1)} - \frac{1}{(s+2)} \\ \frac{-2}{(s+1)} + \frac{2}{(s+2)} & \frac{-1}{(s+1)} + \frac{2}{(s+2)} \end{bmatrix}\end{aligned}$$



Laplace transform Approach to the solution of Homogeneous Equations

$$\Phi(s) = \begin{bmatrix} \frac{2}{(s+1)} - \frac{1}{(s+2)} & \frac{1}{(s+1)} - \frac{1}{(s+2)} \\ \frac{-2}{(s+1)} + \frac{2}{(s+2)} & \frac{-1}{(s+1)} + \frac{2}{(s+2)} \end{bmatrix}$$



$$\Phi(t) = \begin{bmatrix} 2e^{-t} - e^{-2t} & e^{-t} - e^{-2t} \\ -2e^{-t} + 2e^{-2t} & -e^{-t} + 2e^{-2t} \end{bmatrix}$$



Laplace transform Approach to the solution of Homogeneous Equations

Example: Obtain the Homogeneous solution of the following system:

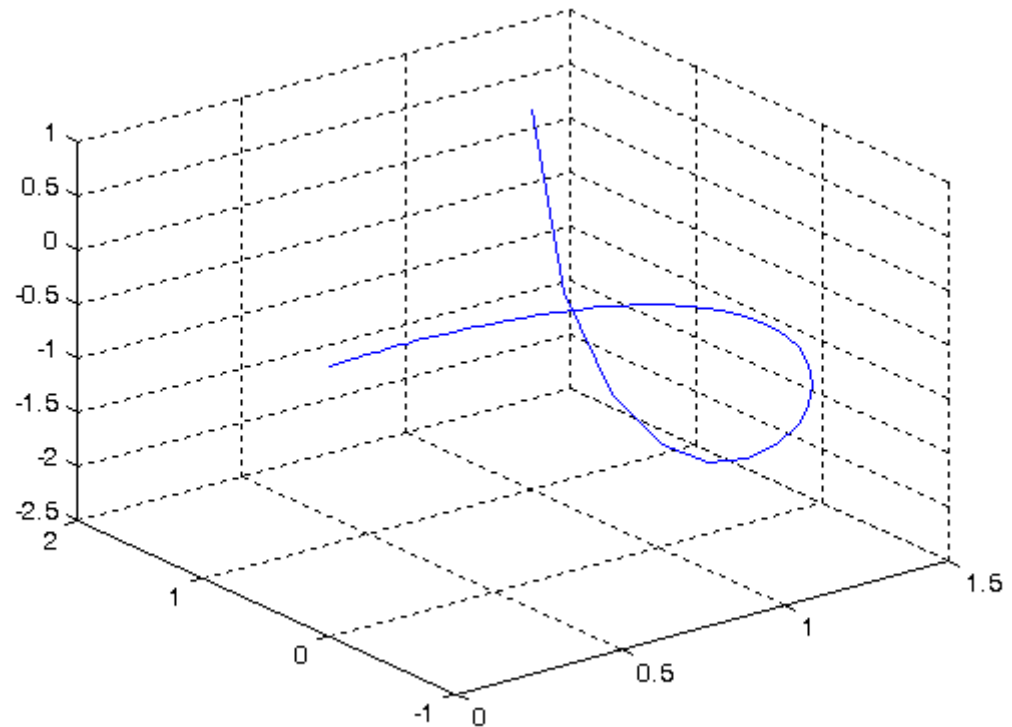
$$A = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ -6 & -11 & -6 \end{bmatrix}$$

$$x(0) = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$



Laplace transform Approach to the solution of Homogeneous Equations

```
a=[0 1 0;0 0 1;-6 -11 -6];  
x0=[1;1;1];  
t=0:0.1:10;  
for i=1:length(t)  
x(:,i)=expm(a*t(i))*x0;  
end  
plot3(x(1,:),x(2,:),x(3,:));  
grid on
```



Matrix Transformation Method for solution of e^{At}

几个特殊矩阵的矩阵指数函数

1. 若A为对角矩阵

$$A = \Lambda = \begin{bmatrix} \lambda_1 & & & 0 \\ & \lambda_2 & & \\ & & \ddots & \\ 0 & & & \lambda_n \end{bmatrix} \quad e^{At} = \Phi(t) = \begin{bmatrix} e^{\lambda_1 t} & & & 0 \\ & e^{\lambda_2 t} & & \\ & & \ddots & \\ 0 & & & e^{\lambda_n t} \end{bmatrix}$$



Matrix Transformation Method for solution of e^{At}

2. 若A能通过非奇异变换对角化 $T^{-1}AT = \Lambda$

$$e^{At} = \Phi(t) = T \begin{bmatrix} e^{\lambda_1 t} & & & 0 \\ & e^{\lambda_2 t} & & \\ & & \ddots & \\ 0 & & & e^{\lambda_n t} \end{bmatrix} T^{-1}$$



Matrix Transformation Method for solution of e^{At}

3. 若A为Jordan阵

$$A = J = \begin{bmatrix} \lambda & 1 & & 0 \\ & \lambda & 1 & \\ & & \ddots & 1 \\ 0 & & & \lambda \end{bmatrix} \quad e^{Jt} = \Phi(t) = e^{\lambda t} \begin{bmatrix} 1 & t & \frac{1}{2!}t^2 & \cdots & \frac{1}{(n-1)!}t^{n-1} \\ 0 & 1 & t & \cdots & \frac{1}{(n-2)!}t^{n-2} \\ \cdots & \cdots & \cdots & \cdots & \cdots \\ 0 & 0 & 0 & \cdots & t \\ 0 & 0 & 0 & \cdots & 1 \end{bmatrix}$$



Matrix Transformation Method for solution of e^{At}

1. 根据定义计算—数值解法

$$e^{At} = I + At + \frac{1}{2!} A^2 t^2 + \cdots + \frac{1}{k!} A^k t^k + \cdots$$

例题1 已知

$$A = \begin{bmatrix} 0 & 1 \\ -2 & -3 \end{bmatrix} \quad \text{求} \quad e^{At}$$

2. 变换A为Jordan标准型

(1). 特征根互异

例题2 采用线性变换法求

$$T^{-1}AT = \Lambda \quad e^{At} = \Phi(t) = T \begin{bmatrix} e^{\lambda_1 t} & & 0 \\ & e^{\lambda_2 t} & \\ & & \ddots \\ 0 & & & e^{\lambda_n t} \end{bmatrix} T^{-1}$$



Matrix Transformation Method for solution of e^{At}

3. 变换A为Jordan标准型

(2). 特征根重根

$$T^{-1}AT = J \quad e^{At} = Te^{Jt}T^{-1}$$

例题2 采用线性
变换法求

$$e^{At} \quad A = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 2 & -5 & 4 \end{bmatrix}$$

4. 采用拉氏反变换法

可以证明 $e^{At} = \Phi(t) = L^{-1}[(sI - A)^{-1}]$

例题3 A阵同例1，采用拉氏反变换法求 e^{At}



Matrix Transformation Method for solution of e^{At}

5. 应用凯莱-哈密顿定理

$$e^{At} = a_{n-1}(t)A^{n-1} + a_{n-2}(t)A^{n-2} + \dots + a_1(t)A + a_0I$$

Case1

特征根互异

Case2

特征根
有重根

$$\begin{bmatrix} a_0(t) \\ a_1(t) \\ \dots \\ a_{n-1}(t) \end{bmatrix} = \begin{bmatrix} 1 & \lambda_1 & \lambda_1^2 & \dots & \lambda_1^{n-1} \\ 1 & \lambda_2 & \lambda_2^2 & \dots & \lambda_2^{n-1} \\ \dots & \dots & \dots & \dots & \dots \\ 1 & \lambda_n & \lambda_n^2 & \dots & \lambda_n^{n-1} \end{bmatrix}^{-1} \begin{bmatrix} e^{\lambda_1 t} \\ e^{\lambda_2 t} \\ \dots \\ e^{\lambda_n t} \end{bmatrix}$$

$$\begin{bmatrix} a_0(t) \\ a_1(t) \\ \dots \\ \dots \\ a_{n-2}(t) \\ a_{n-1}(t) \end{bmatrix} = \begin{bmatrix} 0 & 0 & 0 & \dots & 0 & 1 \\ 0 & 0 & 0 & \dots & 1 & (n-1)\lambda_1 \\ \dots & \dots & \dots & \dots & \dots & \dots \\ 0 & 0 & 1 & \dots & \dots & \frac{(n-1)(n-2)}{2!} \lambda_1^{n-3} \\ 0 & 1 & 2\lambda_1 & \dots & \dots & (n-1)\lambda_1^{n-2} \\ 1 & \lambda_1 & \lambda_1^2 & \dots & \lambda_1^{n-2} & \lambda_1^{n-1} \end{bmatrix}^{-1} \begin{bmatrix} \frac{1}{(n-1)!} t^{n-1} e^{\lambda_1 t} \\ \frac{1}{(n-2)!} t^{n-2} e^{\lambda_1 t} \\ \dots \\ \dots \\ te^{\lambda_1 t} \\ e^{\lambda_1 t} \end{bmatrix}$$



Matrix Transformation Method for solution of e^{At}

例题4 A 阵同例1, 采用凯莱-哈密顿法求 e^{At}

例题5 A 阵同例2, 采用凯莱-哈密顿法求 e^{At}

$$A = \begin{bmatrix} 0 & 1 \\ -2 & -3 \end{bmatrix}$$



Laplace transform Approach to the solution of Homogeneous Equations

Example: A given system

$$\dot{x} = Ax + Bu$$

To satisfy

$$u = 0, x(0) = \begin{bmatrix} 1 \\ -1 \end{bmatrix}, \quad x(t) = \begin{bmatrix} e^{-2t} \\ -e^{-2t} \end{bmatrix}$$

$$u = 0, x(0) = \begin{bmatrix} 2 \\ -1 \end{bmatrix}, \quad x(t) = \begin{bmatrix} 2e^{-t} \\ -e^{-t} \end{bmatrix}$$

Find the system transition matrix $\Phi(t)$ and system matrix A

