

**Department of Systems Engineering**  

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**George Mason University**

**SYST611: Systems  
Methodology and Modeling #1**

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# Outline

- **Introduction**
- **Course Overview and Prerequisites**
- **System Definitions and Concepts**
- **Taxonomy on Models and Methods**
- **Mathematical Models for Dynamical Systems**
- **Solution Properties and Fundamentals**
  - **Differential and Difference Equations**
  - **Matrix Algebra**

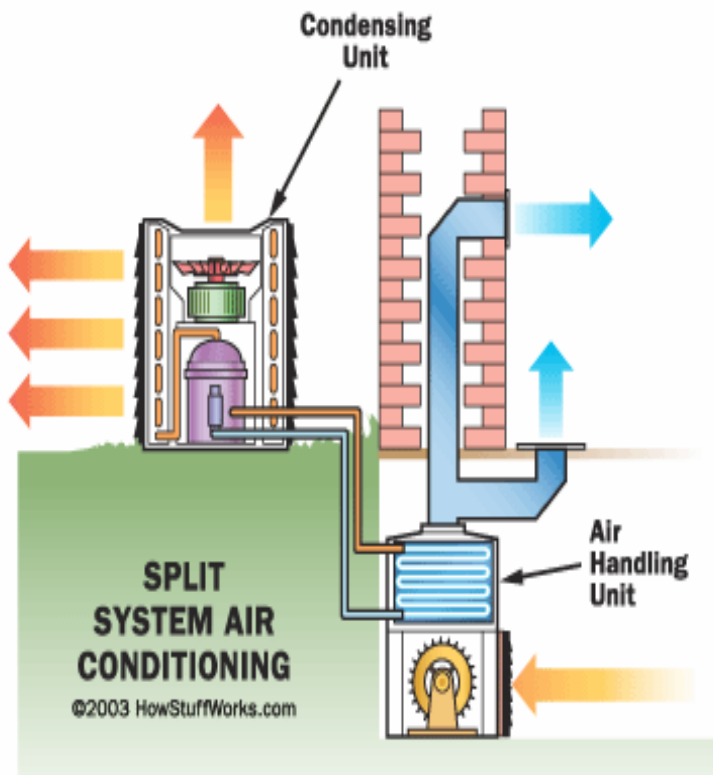
# System Definitions and Concepts

- **A system is a set of interconnected components working together toward some common objectives**
  - Agrees with our intuitive understanding of a *physical system* such as an electrical circuit, a computer system, a transportation system, and a communication system
  - When there is change in time, it's called a *physical dynamical system (PDS)*.
- **An ordered and comprehensive assemblage of facts, principles, or rules**
  - Agrees with the concept of a system as a mathematical abstraction
  - Serves as a model for real system phenomena

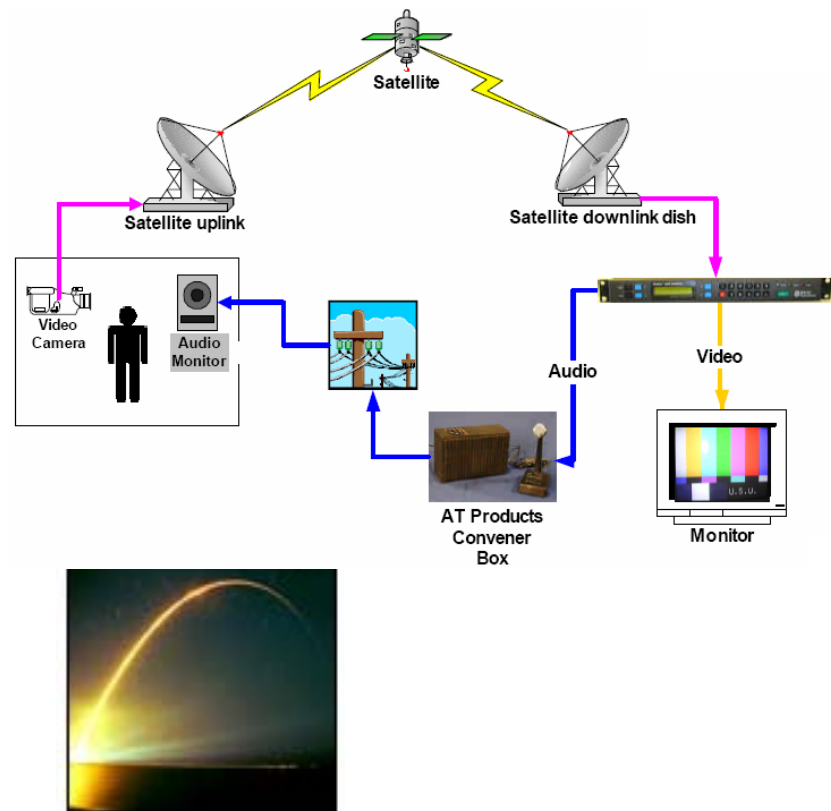
# Systems Engineering

- **Mission: to design PDS's that behave in a desired manner**
  - understand the physical laws
  - develop mathematical models
  - study system behavior
  - system design and implementation
- **Systems Engineering Core Courses**
  - SYST 510: Systems Definition and Cost Modeling
  - SYST 520: System Design and Integration
  - SYST 530: System Management and Evaluation
- **SYST 611: Mathematical foundation for modeling system behavior and optimal performance**
  - required basic method course for all tracks
  - prerequisite: SYST 500 or equivalent - differential/difference equations, matrix algebra, applied probability

# Examples

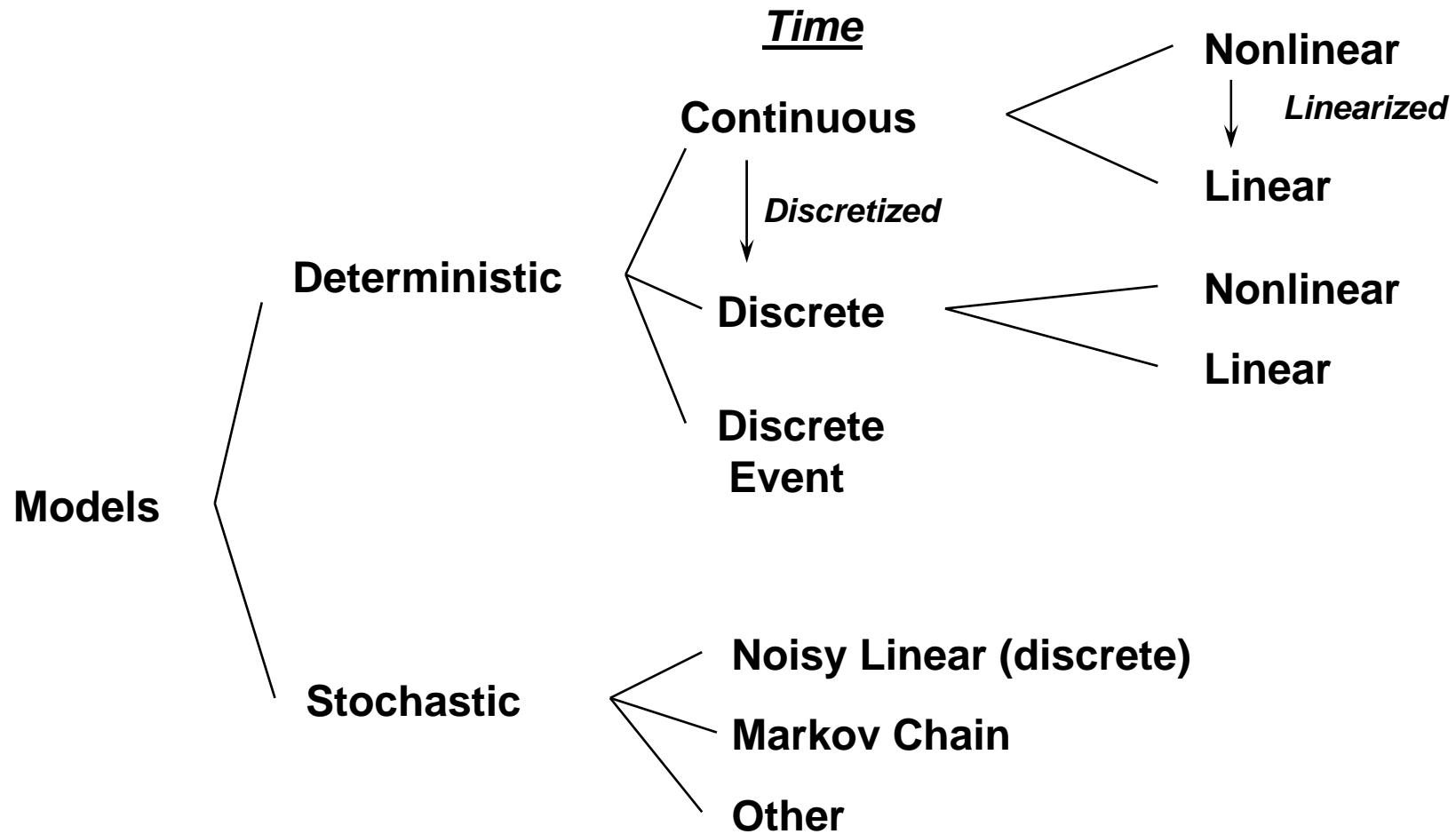


**Air Conditioning System**

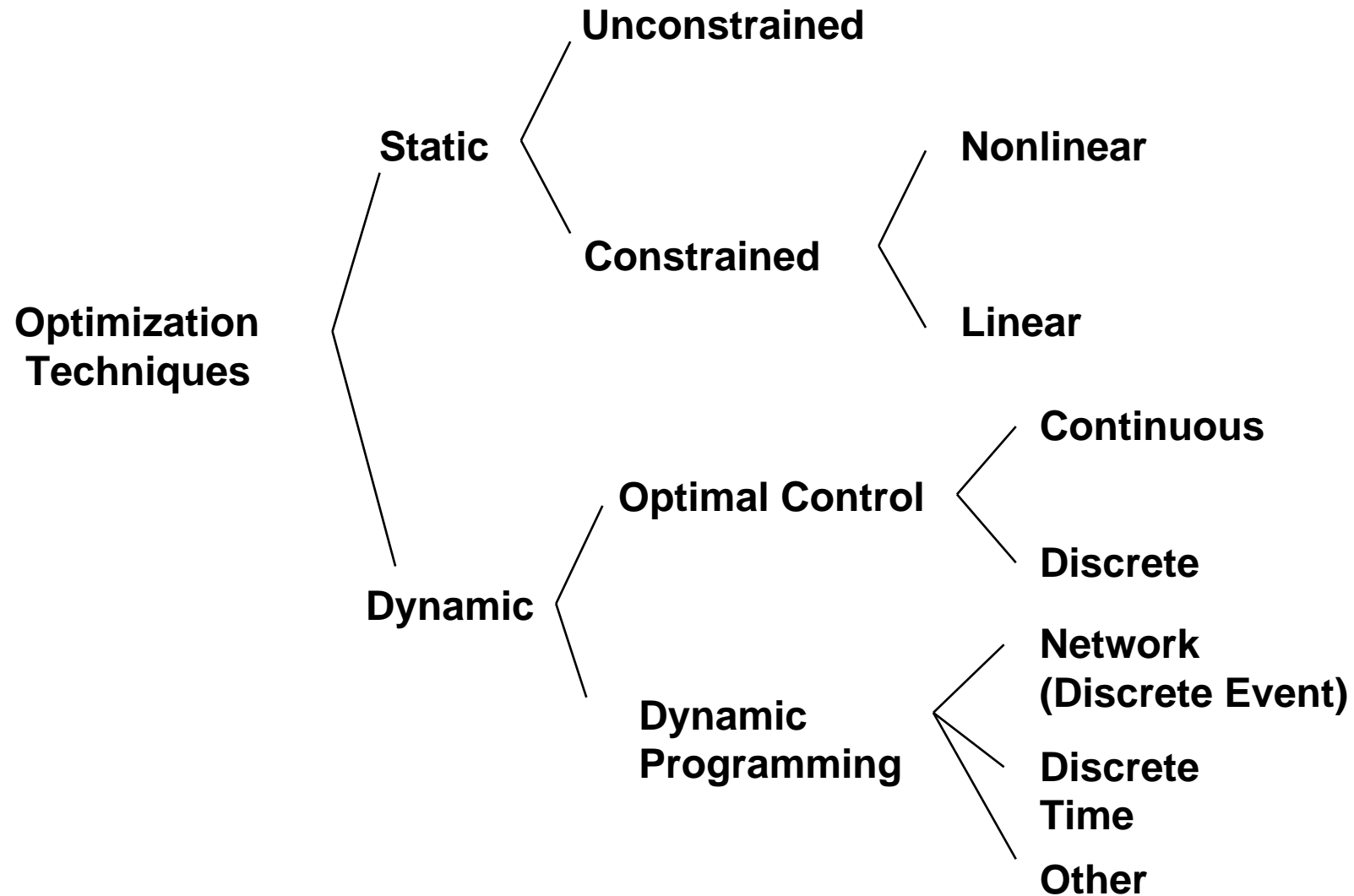


**Satellite System**

# Taxonomy of Models



# Taxonomy of Optimization



# Mathematical Models for Dynamical Systems

- **Model Uncertainties**

- **Deterministic models**: knowledge of the state at time  $t_0$  and the input for  $(t_0, t_1)$  yields the state and the output at time  $t_1$
- **Stochastic models**: knowledge of the state at time  $t_0$  and the input for  $(t_0, t_1)$  yields a probabilistic description of the state and the output at time  $t_1$

- **Time**

- **Continuous-time system**: defined over a continuous time interval
- **Discrete-time system**: defined only over a discrete time set

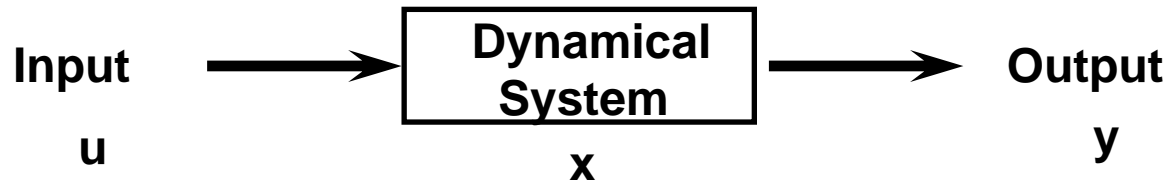
- **Mathematical Form**

- **Linear**: approximations over operating range
- **Nonlinear**: almost all physical systems are nonlinear

# State Representation

- **System I/O Behavior**

- Inputs: external forces
- Outputs: observable measures of the resulting behavior



- **State of a Dynamical System**

- A set of minimum number of variables
- Knowledge of the value of this set of variables at a time  $t_0$ , and of the forcing functions from  $t_0$  to  $t_1$ , where  $t_1 > t_0$ , is sufficient information to determine the output of this system at any time  $t_1 > t_0$

- **Mathematical Model of a PDS**

- Describe the inter-relationship between the state, the input, and the output
- Independent variable: time, discrete or continuous

# Examples

- **Deterministic Models**

- **Geometric growth (discrete time):**

$$x(k+1) = ax(k) \Rightarrow x(k) = a^k x(0)$$

$$x(k+1) = ax(k) + b \Rightarrow x(k) = a^k x(0) + \sum_{i=0}^{k-1} a^{k-1-i} b$$

- **Exponential growth (continuous time):**

$$\frac{dx(t)}{dt} = rx(t) \Rightarrow x(t) = e^{rt} x(0)$$

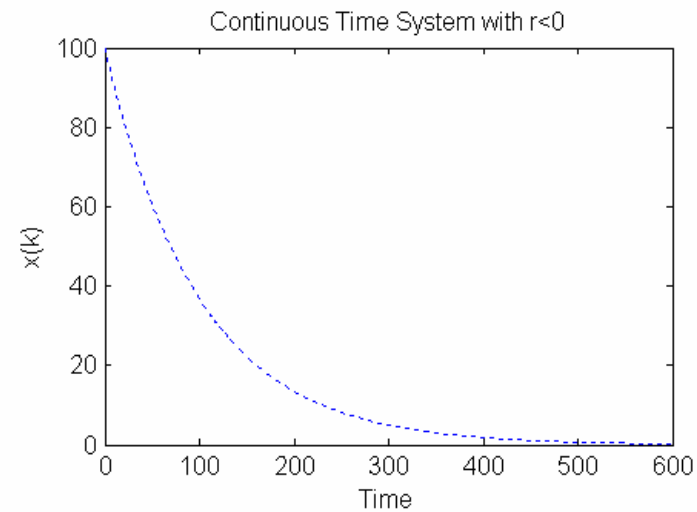
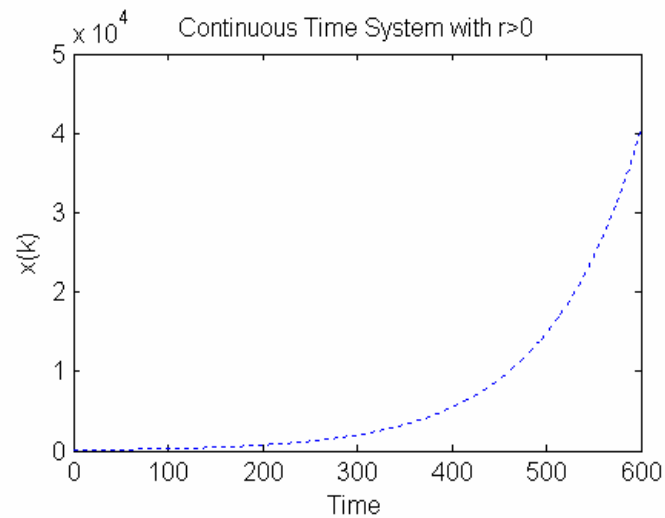
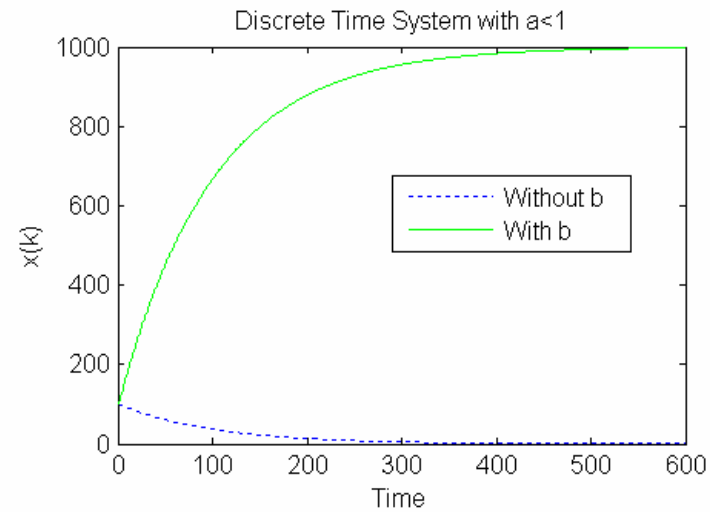
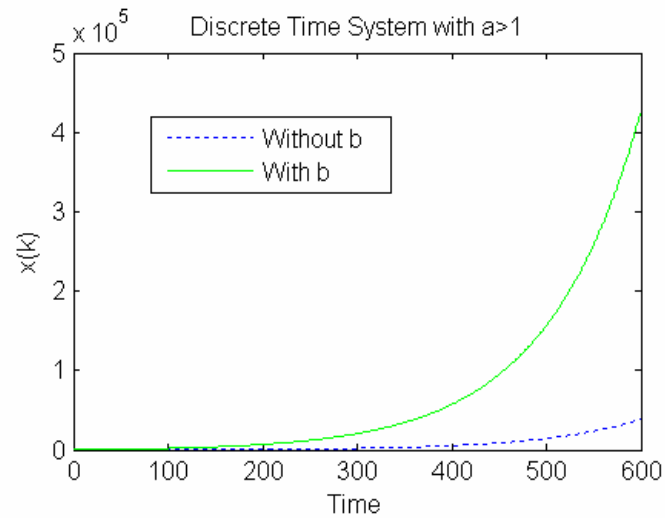
- **Stochastic Models**

- **Random walk (discrete time)**

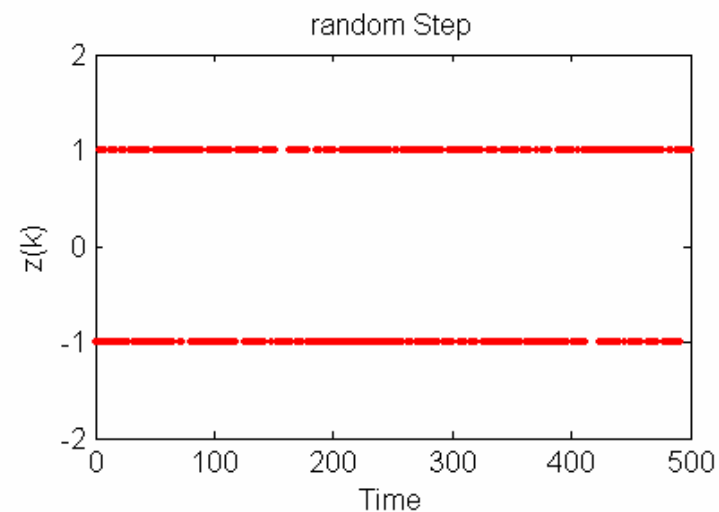
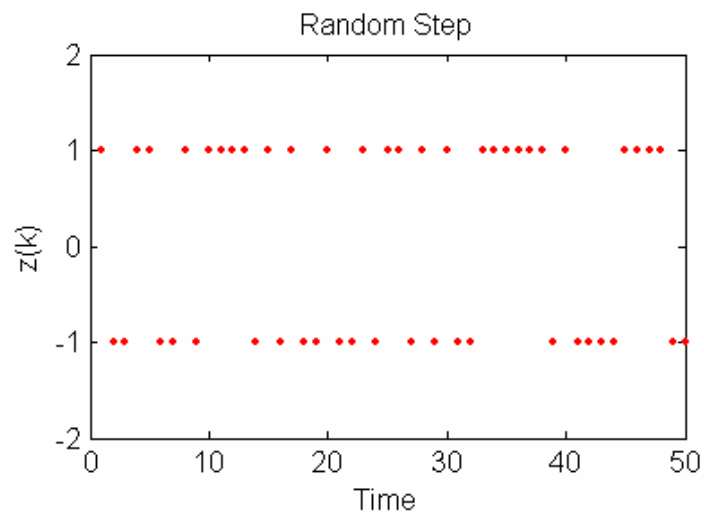
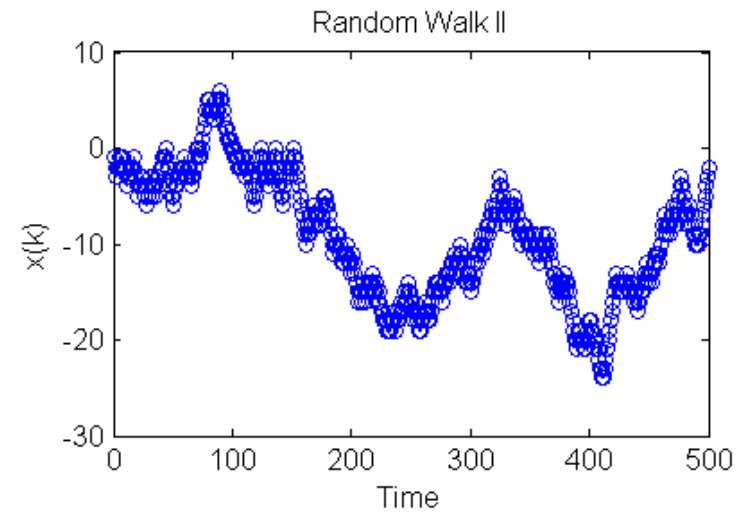
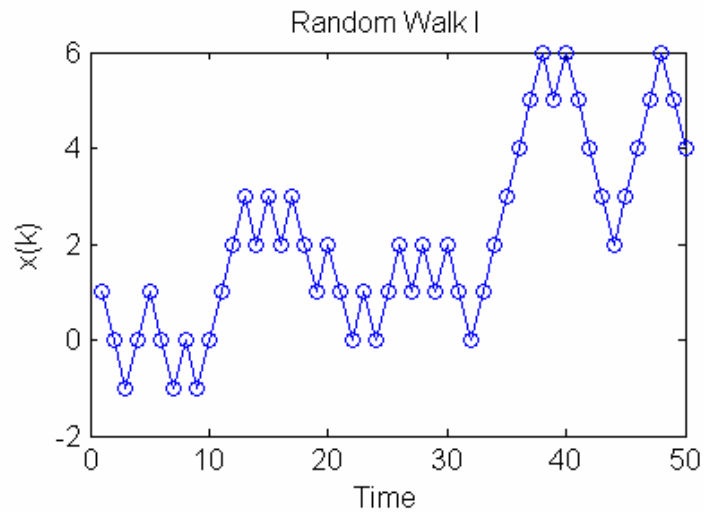
$$x(k) = \sum_{i=1}^k z(i); \quad x(0) = 0, \quad z(k) = \begin{cases} +1 & \text{with } P = 1/2 \\ -1 & \text{with } P = 1/2 \end{cases}$$

$$\Rightarrow P\{x(k) = n\} = \binom{k}{\frac{n+k}{2}} 2^{-k}$$

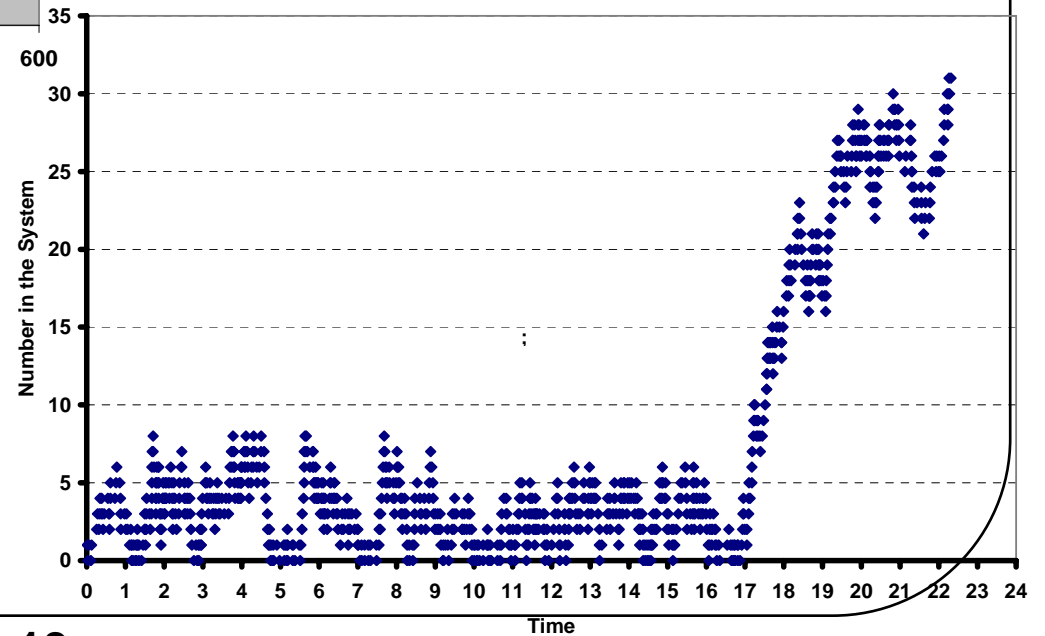
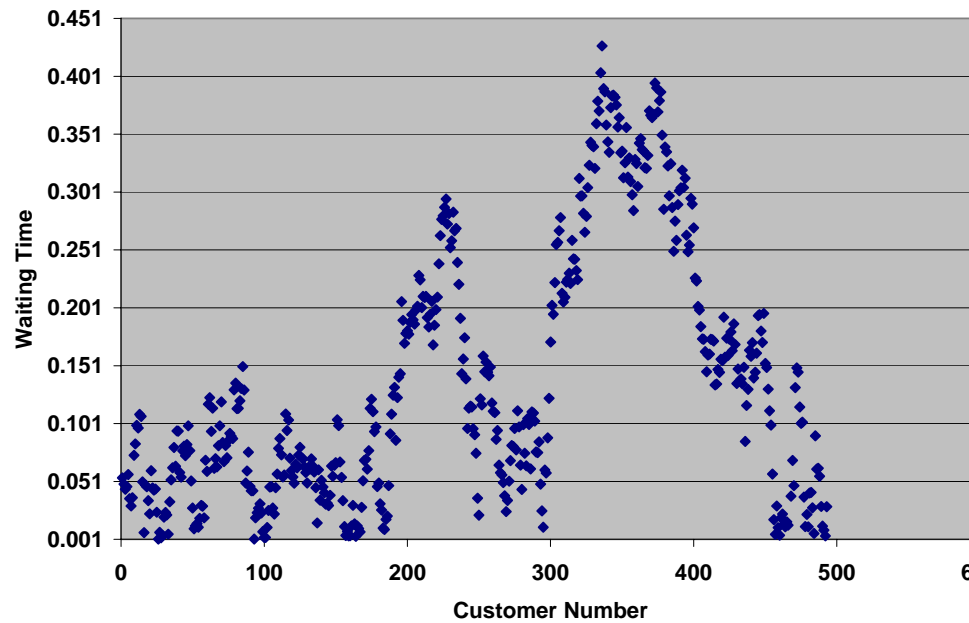
# Deterministic System Examples



# Stochastic System Examples



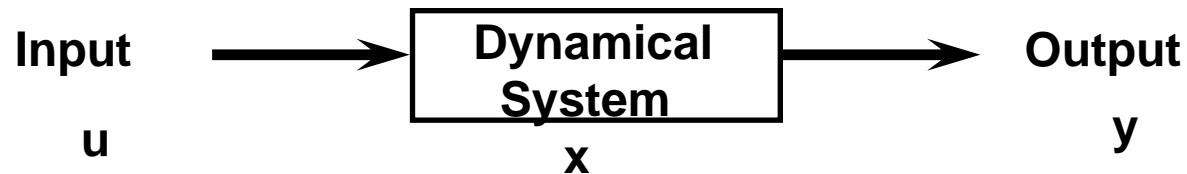
# Stochastic System Example ctd. Queueing Phenomena



# Basic Systems Concept

- **Linearity and Superposition**

- If an input  $u_1(t)$  produces an output  $y_1(t)$  and an input  $u_2(t)$  produces an output  $y_2(t)$ , then input  $c_1u_1(t)+c_2u_2(t)$  produces an output  $c_1y_1(t)+c_2y_2(t)$  for all pairs of inputs  $u_1(t)$  and  $u_2(t)$  and all pairs of constants  $c_1$  and  $c_2$
- It's called the *principle of superposition*.



A linear system obeys the principle of superposition.

- **Causality and Physically Realizable Systems**

- A system is called causal if the output depends only on the present and the past values of the input.
- Causal systems are sometimes called physically realizable systems

# Examples

- **Linear System (*Laplace* transform)**

$$Y_1(s) = L[x_1(t)] = \int x_1(t)e^{-st} dt; \quad Y_2(s) = \int x_2(t)e^{-st} dt$$

$$Y(s) = L[c_1x_1(t) + c_2x_2(t)] = \int [c_1x_1(t) + c_2x_2(t)]e^{-st} dt$$

$$= c_1 \int x_1(t)e^{-st} dt + c_2 \int x_2(t)e^{-st} dt = c_1Y_1(s) + c_2Y_2(s)$$

- **Nonlinear System (Exponential law)**

$$Y_1 = \exp(x_1); \quad Y_2 = \exp(x_2)$$

$$Y = \exp(c_1x_1 + c_2x_2) = \exp(c_1x_1) \times \exp(c_2x_2)$$

$$\neq c_1 \exp(x_1) + c_2 \exp(x_2) = c_1Y_1 + c_2Y_2$$

# Nonlinear Systems: Exponential and Power Law

- **Exponential:**  $y = a (n^{bx})$  or  $y = a (n^{-bx})$ 
  - In differential equations, most commonly  $n = e$  so  $y = ae^{bx}$  or  $y = ae^{-bx}$
  - Common example, population growth
    - » Population growth:  $N(n+1) = N(n) * k$ ,
    - »  $N(n) = N(0) * k^n$ , which is in the form  $y(x) = an^x$  (where  $n = x$ ,  $k = a$ ), and is therefore exponential
    - » Key point about exponentials: the amount of increase or decrease depends on how much you currently have; this phenomena occurs often in nature
- **Power Law:**  $y = ax^{bn}$  or  $y = ax^{-bn}$ 
  - Does not grow (or decay) as fast as an exponential
  - Occurs when “large is rare and small is common,” i.e.
    - » Frequency of earthquakes (large earthquakes rare, small are common)
    - » Number of connection (degree) in a complex network (internet, WWW, etc.)

# Issues in Modeling

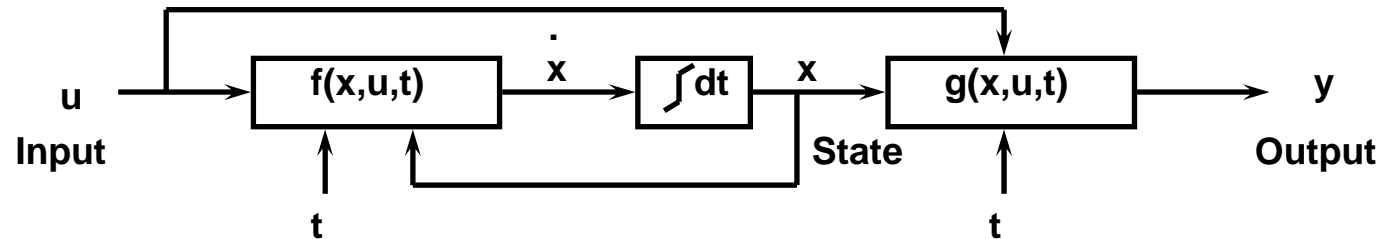
- **System Description Models**
  - Linear systems
  - Nonlinear systems
  - Linearization
- **System Time Models**
  - Discrete time
  - Continuous time
  - Discretization
- **System Behavior Models**
  - Input-output relationship
  - Stability
  - Controllability

# System Equations

- **Continuous-Time Systems**

- Ordinary differential equations

$$\begin{aligned} \dot{x}(t) &= f(x, u, t) \\ y &= g(x, u, t) \end{aligned}$$



- **Discrete-Time Systems**

- Difference equations

$$\begin{aligned} x_{k+1} &= f(x_k, u_k, k) \\ y_k &= g(x_k, u_k, k) \end{aligned}$$

# Examples – Discrete Time

## • National Economics

$$Y(k) = C(k) + I(k) + G(k)$$

$$C(k) = mY(k)$$

$$Y(k+1) - Y(k) = rI(k)$$

$$\Rightarrow Y(k+1) = [1 + r(1 - m)]Y(k) - rG(k)$$

**Y, C, I, G** represent national income/product, consumption, investment, and government expenditure respectively

**m**: portion of income to consume, **r**: growth factor

## • A Migration Model

$$r(k+1) = \alpha r(k) - \beta \{r(k) - \gamma [r(k) + u(k)]\}$$

$$u(k+1) = \alpha u(k) + \beta \{r(k) - \gamma [r(k) + u(k)]\}$$

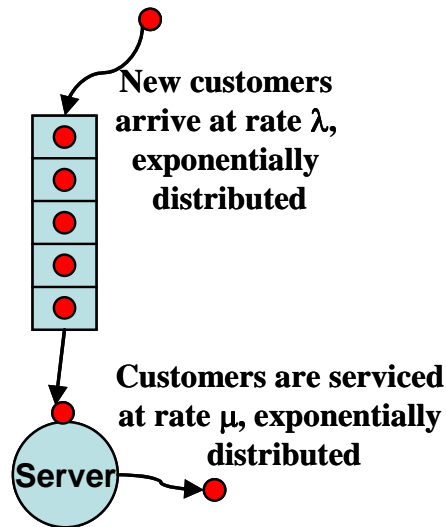
$$\begin{bmatrix} r(k+1) \\ u(k+1) \end{bmatrix} = \begin{bmatrix} \alpha - \beta(1 - \gamma) & \beta\gamma \\ \beta(1 - \gamma) & \alpha - \beta\gamma \end{bmatrix} \begin{bmatrix} r(k) \\ u(k) \end{bmatrix}$$

**r** and **u** represent the rural and urban populations

**$\alpha$** : yearly growth factor,  **$\beta$** : migration factor

**$\gamma(r+u)$** : optimal rural population

# Discrete Time, ctd. Queueing Phenomena



PDF  $a(t) = \lambda e^{-\lambda t}$  describes *interarrival time Probability distribution*

PDF  $b(t) = \mu e^{-\mu t}$  describes *service time probability distribution*

- Let  $N$  = number of customers in the system,  $\Pr(N)$  is the probability that there are  $N$  customers in the system, then flow balance equations show
- $(\lambda + \mu) \Pr(N) = \lambda \Pr(N - 1) + \mu \Pr(N + 1)$
- Solving recursively yields  $\Pr(N) = (1 - \lambda/\mu) (\lambda/\mu)^N$

# Examples – Continuous Time

- **Richardson's Theory of arm race**

$$\dot{x}(t) = ky(t) - \alpha x(t) + g$$

$$\dot{y}(t) = lx(t) - \beta y(t) + h$$

or

$$\begin{bmatrix} \dot{x}(t) \\ \dot{y}(t) \end{bmatrix} = \begin{bmatrix} -\alpha & k \\ l & -\beta \end{bmatrix} \begin{bmatrix} x(t) \\ y(t) \end{bmatrix} + \begin{bmatrix} g \\ h \end{bmatrix}$$

**x and y represent the arm level of two competing nations**

**g and h are "grievances" factor (revenge)**

**K and l are "defense" coefficients,  $\alpha, \beta$  are "fatigue" coefficients**

- **The Predator-Prey model (Lotka-Volterra eqn.)**

$$\dot{N}_1(t) = aN_1 - bN_1N_2$$

$$\dot{N}_2(t) = -cN_2 + dN_1N_2$$

**N1 and N2 represent the prey and predator populations**

**a and c are natural growth and death rate of prey and predator**

**b and d are "encounters" coefficients**

# Linear Systems

- **Linear System Representation**

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Bu(t); & x(0) &= x_0 \\ y(t) &= Cx(t) + Du(t)\end{aligned}$$

$$\begin{aligned}x_{k+1} &= Fx_k + Gu_k; & x_0 &\text{ is known} \\ y_k &= Hx_k + Ju_k\end{aligned}$$

# Linear System Solutions

## Discrete Time:

$$x_{k+1} = Fx_k + Gu_k; x_0 \text{ given}$$

$$\text{General solution: } x_k = x_{h,k} + x_{p,k}$$

**Homogeneous solution** : *unforced equation with an initial condition*

$$x_{h,k+1} = Fx_{h,k}; x_{h,0} = x_0$$

$$\Rightarrow x_{h,k} = F^k x_0$$

**Particular solution** : *forced equation with a zero initial condition*

$$x_{p,k+1} = Fx_{p,k} + Gu_k; x_{p,0} = 0$$

$$\Rightarrow x_{p,k} = \sum_{i=0}^{k-1} F^{k-1-i} Gu_i$$

**Total solution** :

$$x_k = F^k x_0 + \sum_{i=0}^{k-1} F^{k-1-i} Gu_i$$

# Linear System Solutions

## Continuous Time:

$$\dot{x}(t) = Ax(t) + Bu(t); \quad x(0) = x_0$$

*General solution:*  $x(t) = x_h(t) + x_p(t)$

**Homogeneous solution:** *unforced equation with an initial condition*

$$\dot{x}_h(t) = Ax_h(t); \quad x_h(0) = x_0$$

$$\Rightarrow x_h(t) = e^{At} x_0$$

**Particular solution:** *forced equation with a zero initial condition*

$$\dot{x}_p(t) = Ax_p(t) + Bu(t); \quad x_p(0) = 0$$

$$\Rightarrow x_p(t) = \int_0^t e^{A(t-\tau)} Bu(\tau) d\tau$$

**Total solution:**  $x(t) = e^{At} x_0 + \int_0^t e^{A(t-\tau)} Bu(\tau) d\tau$

where  $e^{At} = I + At + \frac{(At)^2}{2} + \frac{(At)^3}{6} + \dots = \sum_{i=0}^{\infty} \frac{(At)^i}{i!}$

# Matrix Algebra

For an  $n \times n$  square matrix  $A$ , if the number  $\lambda$  and nonzero vector  $u$  satisfy  $Au = \lambda u$  or  $(A - \lambda I)u = 0$  then  $\lambda$  is called an *eigenvalue* of  $A$  and  $u$  is the corresponding *eigenvector*

A non - zero solution  $u$  exists only if  $\det[A - \lambda I] = 0$ , but

$$\det[A - \lambda I] = \det \begin{bmatrix} a_{11} - \lambda & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} - \lambda & \dots & a_{2n} \\ \cdot & \cdot & \cdot & \cdot \\ a_{n1} & a_{n2} & \dots & a_{nn} - \lambda \end{bmatrix}$$
$$= \lambda^n + p_1 \lambda^{n-1} + \dots + p_{n-1} \lambda + p_n = p(\lambda)$$

If  $\lambda_1, \lambda_2, \dots, \lambda_n$  are the solutions of the Characteristic Polynomial  $p(\lambda)$ , then  $p(\lambda) = (\lambda - \lambda_1)(\lambda - \lambda_2) \dots (\lambda - \lambda_n)$  or  $p(\lambda_i) = 0$

Cayley – Hamilton Theorem

$$p(A) = A^n + p_1 A^{n-1} + \dots + p_{n-1} A + p_n = 0$$

# Matrix Algebra

Assuming  $A$  has  $n$  distinct *eigenvalues*  $\lambda_1, \lambda_2, \dots, \lambda_n$ , and that the corresponding *eigenvectors* are  $u_1, u_2, \dots, u_n$ , Then the eigenvectors are linearly independent and can be used as the columns of a non-singular matrix  $T$

$$\begin{aligned}AT &= A[u_1 \ u_2 \ \dots \ u_n] = [Au_1 \ Au_2 \ \dots \ Au_n] \\&= [\lambda_1 u_1 \ \lambda_2 u_2 \ \dots \ \lambda_n u_n] = T\Lambda \quad \text{where } \Lambda = \text{diag}[\lambda_i] \\&\Rightarrow T^{-1}AT = \Lambda, \quad A = T\Lambda T^{-1} \\A^k &= AA \cdots A = T\Lambda T^{-1} T\Lambda T^{-1} \cdots T\Lambda T^{-1} = T\Lambda^k T^{-1} \\&\Rightarrow p(A) = Tp(\Lambda)T^{-1}, \quad \text{and} \quad e^{At} = Te^{\Lambda t}T^{-1}\end{aligned}$$

- Therefore, both the polynomial and exponential functions of a matrix can be represented as functions of eigenvalues and eigenvectors of the matrix
- Solution properties of a system can then be determined

# Matrix Example

$$A = \begin{bmatrix} 2 & 1 \\ 0 & 3 \end{bmatrix}, \quad Q: A^k = ? \quad e^{At} = ?$$

$$(A - \lambda I)\mu = 0 \Rightarrow \lambda = 2, 3; \quad \mu = \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

$$\begin{aligned} A^k &= T\Lambda^k T^{-1} = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 2 & 0 \\ 0 & 3 \end{bmatrix}^k \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}^{-1} \\ &= \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 2^k & 0 \\ 0 & 3^k \end{bmatrix} \begin{bmatrix} 1 & -1 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} 2^k & 3^k - 2^k \\ 0 & 3^k \end{bmatrix} \end{aligned}$$

$$\begin{aligned} e^{At} &= T e^{\Lambda t} T^{-1} = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} e^{\begin{bmatrix} 2 & 0 \\ 0 & 3 \end{bmatrix} t} \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}^{-1} \\ &= \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} e^{2t} & 0 \\ 0 & e^{3t} \end{bmatrix} \begin{bmatrix} 1 & -1 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} e^{2t} & e^{3t} - e^{2t} \\ 0 & e^{3t} \end{bmatrix} \end{aligned}$$

# Linear System Example

**F r e e F a l l b o d y :**

$$\ddot{y} = \frac{d^2 y}{dt^2} = g, \quad \dot{y} = g t + v_0$$
$$\Rightarrow y(t) = \frac{1}{2} g t^2 + v_0 t + y_0$$

Let  $x_1 = y$  and  $\dot{x}_1 = dx_1 / dt = x_2$ , then  $\dot{x}_2 = \ddot{x}_1 = g$   
i.e.,

$$\begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \dot{x}(t) = Ax(t) + Bu = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} 0 \\ 1 \end{bmatrix} g$$

but  $A^k = 0$  for  $k \geq 2$ ,  $e^{At} = I + At = \begin{bmatrix} 1 & t \\ 0 & 1 \end{bmatrix}$ , with  $x(0) = \begin{bmatrix} y_0 \\ v_0 \end{bmatrix}$

# System Solution

*Solution*

$$\begin{aligned}\begin{bmatrix} x_1(t) \\ x_2(t) \end{bmatrix} &= e^{At} x_0 + \int_0^t e^{A(t-\tau)} B u(\tau) d\tau \\ &= \begin{bmatrix} 1 & t \\ 0 & 1 \end{bmatrix} \begin{bmatrix} y_0 \\ v_0 \end{bmatrix} + \int_0^t \begin{bmatrix} 1 & (t-\tau) \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 0 \\ g \end{bmatrix} d\tau \\ &= \begin{bmatrix} y_0 + v_0 t + \frac{1}{2} g t^2 \\ v_0 + g t \end{bmatrix} = \begin{bmatrix} y(t) \\ \dot{y}(t) \end{bmatrix}\end{aligned}$$