

Time Response of Dynamic Systems

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Multi-dimensional trajectories
Numerical integration
Linear and nonlinear systems
Linearization of nonlinear models
LTI System Response
Phase-plane plots

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Multi-Dimensional Trajectories

Position, velocity, and acceleration are vectors

$$\mathbf{x} = \begin{bmatrix} x \\ y \end{bmatrix}; \quad \mathbf{v} = \begin{bmatrix} v_x \\ v_y \end{bmatrix}; \quad \mathbf{a} = \begin{bmatrix} a_x \\ a_y \end{bmatrix}; \quad \mathbf{j} = \begin{bmatrix} j_x \\ j_y \end{bmatrix}; \quad \mathbf{s} = \begin{bmatrix} s_x \\ s_y \end{bmatrix}$$

$$\begin{bmatrix} x(0) \\ x(t) \\ v_x(0) \\ v_x(t) \\ a_x(0) \\ a_x(t) \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 & 0 \\ 1 & t & t^2/2 & t^3/6 & t^4/24 & t^5/120 \\ 0 & 1 & 0 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} x(0) \\ v_x(0) \\ a(0) \end{bmatrix}$$

$$\begin{bmatrix} y(0) \\ y(t) \\ v_y(0) \\ v_y(t) \\ a_y(0) \\ a_y(t) \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 & 0 \\ 1 & t & t^2/2 & t^3/6 & t^4/24 & t^5/120 \\ 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & t & t^2/2 & t^3/6 & t^4/24 \\ 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & t & t^2/2 & t^3/6 \end{bmatrix} \begin{bmatrix} y(0) \\ v_y(0) \\ a_y(0) \\ j_y(0) \\ s_y(0) \\ c_y(0) \end{bmatrix}$$

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Two-Dimensional Trajectory

Solve for Cartesian components separately

x Component

$$\begin{bmatrix} j_x(0) \\ s_x(0) \\ c_x(0) \end{bmatrix} = \begin{bmatrix} -60/t^3 & 60/t^3 & -36/t^2 & -24/t^2 & -9/t & 3/t \\ 360/t^4 & -360/t^4 & 192/t^3 & 168/t^3 & 36/t^2 & -24/t^2 \\ -720/t^5 & 720/t^5 & -360/t^4 & -360/t^4 & -60/t^3 & 60/t^3 \end{bmatrix} \begin{bmatrix} x(0) \\ x(t) \\ v_x(0) \\ v_x(t) \\ a_x(0) \\ a_x(t) \end{bmatrix}$$

y Component

$$\begin{bmatrix} j_y(0) \\ s_y(0) \\ c_y(0) \end{bmatrix} = \begin{bmatrix} -60/t^3 & 60/t^3 & -36/t^2 & -24/t^2 & -9/t & 3/t \\ 360/t^4 & -360/t^4 & 192/t^3 & 168/t^3 & 36/t^2 & -24/t^2 \\ -720/t^5 & 720/t^5 & -360/t^4 & -360/t^4 & -60/t^3 & 60/t^3 \end{bmatrix} \begin{bmatrix} y(0) \\ y(t) \\ v_y(0) \\ v_y(t) \\ a_y(0) \\ a_y(t) \end{bmatrix}$$

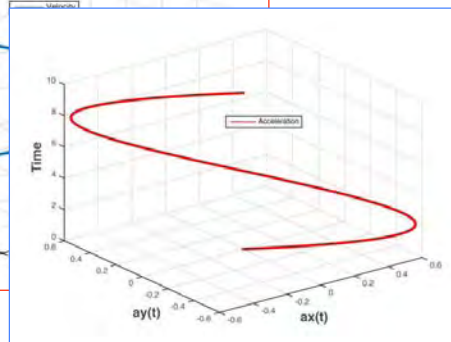
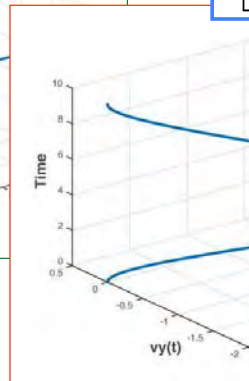
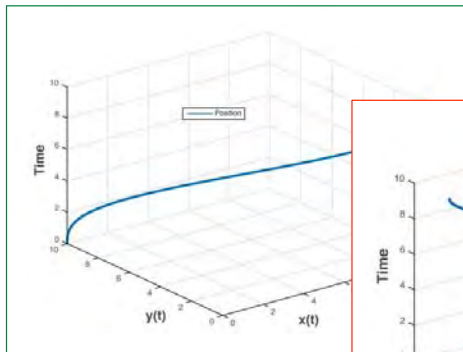
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Two-Dimensional Example

Required acceleration vector is specified by

$$\mathbf{a}(t) = \mathbf{a}(0) + \mathbf{j}(0)t + \mathbf{s}(0)t^2/2 + \mathbf{c}t^3/6$$

$$\begin{aligned} &= \mathbf{a}_{control}(t) + \mathbf{a}_{gravity}(t) + \mathbf{a}_{disturbance}(t) \\ &= [\mathbf{f}_{control}(t) + \mathbf{f}_{gravity}(t) + \mathbf{f}_{disturbance}(t)] / m(t) \end{aligned}$$



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Six-Degree-of-Freedom (Rigid Body) Equations of Motion

$$\begin{aligned}\dot{\mathbf{r}}_I &= \mathbf{H}_B^I \mathbf{v}_B \\ \dot{\mathbf{v}}_B &= \frac{1}{m} \mathbf{f}_B - \tilde{\boldsymbol{\omega}}_B \mathbf{v}_B\end{aligned}$$

Translational position and velocity

$$\mathbf{r}_I = \begin{bmatrix} x \\ y \\ z \end{bmatrix}; \quad \mathbf{v}_B = \begin{bmatrix} u \\ v \\ w \end{bmatrix}$$

Rotational position and velocity

$$\boldsymbol{\Theta} = \begin{bmatrix} \phi \\ \theta \\ \psi \end{bmatrix}; \quad \boldsymbol{\omega}_B = \begin{bmatrix} p \\ q \\ r \end{bmatrix}$$

$$\begin{aligned}\dot{\boldsymbol{\Theta}} &= \mathbf{L}_B^I \boldsymbol{\omega}_B \\ \dot{\boldsymbol{\omega}}_B &= \mathbf{I}_B^{-1} (\mathbf{m}_B - \tilde{\boldsymbol{\omega}}_B \mathbf{I}_B \boldsymbol{\omega}_B)\end{aligned}$$

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... nonlinear, and complex

Rate of change of Translational Position

$$\begin{aligned}\dot{x}_I &= (\cos \theta \cos \psi)u + (-\cos \phi \sin \psi + \sin \phi \sin \theta \cos \psi)v + (\sin \phi \sin \psi + \cos \phi \sin \theta \cos \psi)w \\ \dot{y}_I &= (\cos \theta \sin \psi)u + (\cos \phi \cos \psi + \sin \phi \sin \theta \sin \psi)v + (-\sin \phi \cos \psi + \cos \phi \sin \theta \sin \psi)w \\ \dot{z}_I &= (-\sin \theta)u + (\sin \phi \cos \theta)v + (\cos \phi \cos \theta)w\end{aligned}$$

Rate of change of Translational Velocity

$$\begin{aligned}\dot{u} &= X/m - g \sin \theta + rv - qw \\ \dot{v} &= Y/m + g \sin \phi \cos \theta - ru + pw \\ \dot{w} &= Z/m + g \cos \phi \cos \theta + qu - pv\end{aligned}$$

Rate of change of Angular Position

$$\begin{aligned}\dot{\phi} &= p + (q \sin \phi + r \cos \phi) \tan \theta \\ \dot{\theta} &= q \cos \phi - r \sin \phi \\ \dot{\psi} &= (q \sin \phi + r \cos \phi) \sec \theta\end{aligned}$$

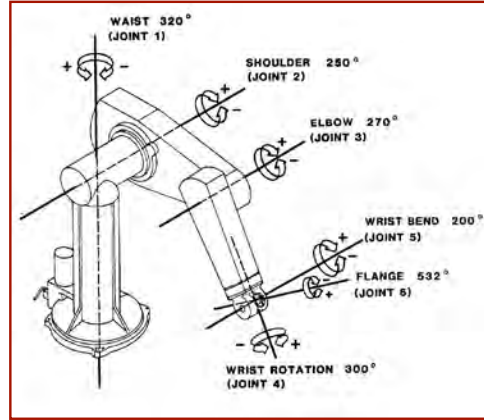
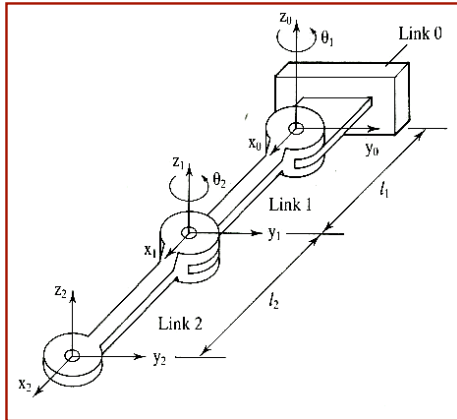
Rate of change of Angular Velocity

$$\begin{aligned}\dot{p} &= (I_{zz}L + I_{xz}N - \{I_{xz}(I_{yy} - I_{xx} - I_{zz})p + [I_{xz}^2 + I_{zz}(I_{zz} - I_{yy})]r\}q) / (I_{xx}I_{zz} - I_{xz}^2) \\ \dot{q} &= [M - (I_{xx} - I_{zz})pr - I_{xz}(p^2 - r^2)] / I_{yy} \\ \dot{r} &= (I_{xz}L + I_{xx}N - \{I_{xz}(I_{yy} - I_{xx} - I_{zz})r + [I_{xz}^2 + I_{xx}(I_{xx} - I_{yy})]p\}q) / (I_{xx}I_{zz} - I_{xz}^2)\end{aligned}$$

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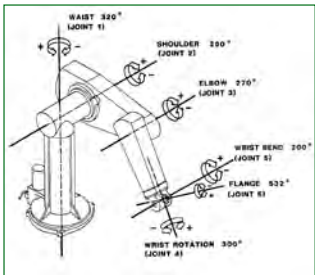
Multiple Rigid Links Lead to Multiple Constraints

- Each link is subject to the same 6-DOF rigid-body dynamic equations
- ... but each link is constrained to have a single degree of freedom w.r.t. proximal link



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Newton-Euler Link Dynamics



- Link dynamics are coupled
 - Proximal-link loads affected by distal-link positions and velocities
 - Distal-link accelerations affected by proximal-link motions
 - Joints produce constraints on link motions

- Net forces and torques at each joint related to velocities and accelerations of the centroids of the links
- Equations of motion derived directly for each link, with constraints

$$\dot{\mathbf{v}}_B = \frac{1}{m} \mathbf{f}_B - \tilde{\boldsymbol{\omega}}_B \mathbf{v}_B$$

$$\dot{\boldsymbol{\omega}}_B = \mathbf{I}_B^{-1} (\mathbf{m}_B - \tilde{\boldsymbol{\omega}}_B \mathbf{I}_B \boldsymbol{\omega}_B)$$

$$\dot{\mathbf{x}}(t) = \mathbf{f}[\mathbf{x}(t), \mathbf{u}(t), \mathbf{w}(t), \mathbf{p}(t), t]$$

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Lagrangian Link Dynamics

- **Lagrange's equation** derives from Newton's Laws
 - Principle of virtual work
 - D'Alembert's principle
- Dynamic behavior described by work done and energy stored in the system
- Equations of motion derived from **Lagrangian** function and Lagrange's equation

Generalized coordinate, q_n , e.g., $(x, y, z, \psi, \theta, \phi)$
 Generalized force, F_n , e.g., $(f_x, f_y, f_z, m_x, m_y, m_z)$

$$L(q_n, \dot{q}_n) \triangleq KE - PE$$

$$\frac{d}{dt} \left(\frac{dL(q_n, \dot{q}_n)}{d\dot{q}_n} \right) - \frac{\partial L(q_n, \dot{q}_n)}{\partial q_n} = F_n$$

$$\rightarrow \dot{\mathbf{x}}(t) = \mathbf{f}[\mathbf{x}(t), \mathbf{u}(t), \mathbf{w}(t), \mathbf{p}(t), t]$$

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Hamiltonian Link Dynamics

- **Hamilton's Principle:** *Lagrange's equation* is a necessary condition for an extremum (i.e., maximum or minimum)

$$\text{extremum } I = \int_{t_1}^{t_2} L(q_n, \dot{q}_n) dt$$

Equations of motion derived from an optimization problem

- **Hamiltonian function**

$$H(p, q) = \sum \dot{q}_n p_n - L(q_n, \dot{q}_n)$$

- **Hamilton's equations**

$$\dot{q}_n = \frac{\partial H(p, q)}{\partial p_n}$$

$$\dot{p}_n = -\frac{\partial H(p, q)}{\partial q_n}$$

Generalized momentum, p_n ,
 e.g., $(mv_x, mv_y, mv_z, I_{xx}p, I_{yy}q, I_{zz}r)$

$$\rightarrow \dot{\mathbf{x}}(t) = \mathbf{f}[\mathbf{x}(t), \mathbf{u}(t), \mathbf{w}(t), \mathbf{p}(t), t]$$

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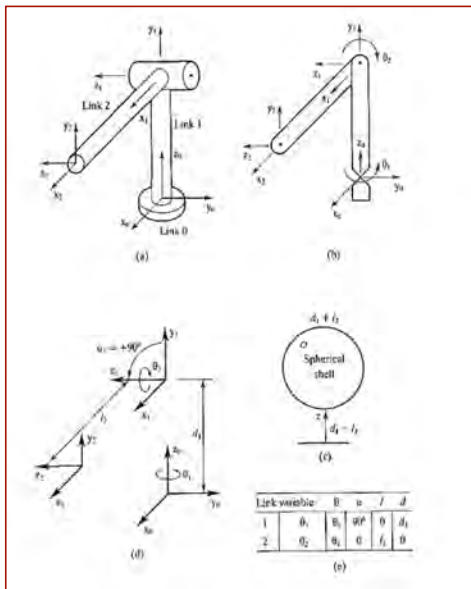
Example: Two-Link Robot Equations of Motion

State vector

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} \text{Angle of 1st link, } \theta_1, \text{ rad} \\ \text{Angular rate of 1st link, } \dot{\theta}_1, \text{ rad/sec} \\ \text{Angle of 2nd link, } \theta_2, \text{ rad} \\ \text{Angular rate of 2nd link, } \dot{\theta}_2, \text{ rad/sec} \end{bmatrix}$$

Control vector

$$\begin{bmatrix} u_1 \\ u_2 \end{bmatrix} = \begin{bmatrix} \tau_1, \text{ torque at 1st joint} \\ \tau_2, \text{ torque at 2nd joint} \end{bmatrix}$$



McKerrow, 1991

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When Possible, Simplify the Equations

Two-link robot equations of motion

- Mass, m , located at end of Link 2
- Inertias of Links 1 and 2 neglected

$$\begin{aligned} \dot{x}_1 &= x_2 \\ \dot{x}_2 &= \frac{1}{\cos^2 x_3} \left(x_2 x_4 \sin 2x_3 + \frac{u_1}{ml_2^2} \right) \\ \dot{x}_3 &= x_4 \\ \dot{x}_4 &= -\frac{g}{l_2} \cos x_3 - \frac{x_2}{2} \sin 2x_3 + \frac{u_2}{ml_2^2} \end{aligned}$$

from Example 7.1, Lagrangian derivation, McKerrow, 1991, pp. 388-390.

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Differential Equations Integrated to Produce Time Response

$$\dot{\mathbf{x}}(t) = \mathbf{f}[\mathbf{x}(t), \mathbf{u}(t), \mathbf{w}(t), \mathbf{p}(t), t], \quad \mathbf{x}(0) \text{ given}$$

$$\mathbf{x}(t) = \mathbf{x}(0) + \int_0^t \mathbf{f}[\mathbf{x}(t), \mathbf{u}(t), \mathbf{w}(t), \mathbf{p}(t), t] dt$$

Numerical integration is an approximation

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Rectangular and Trapezoidal Integration of Differential Equations

Rectangular (Euler) Integration

$$\begin{aligned} \mathbf{x}(t_k) &= \mathbf{x}(t_{k-1}) + \Delta \mathbf{x}(t_{k-1}, t_k) \\ &\approx \mathbf{x}(t_{k-1}) + \mathbf{f}[\mathbf{x}(t_{k-1}), \mathbf{u}(t_{k-1}), \mathbf{w}(t_{k-1})] \Delta t, \quad \Delta t = t_k - t_{k-1} \end{aligned}$$

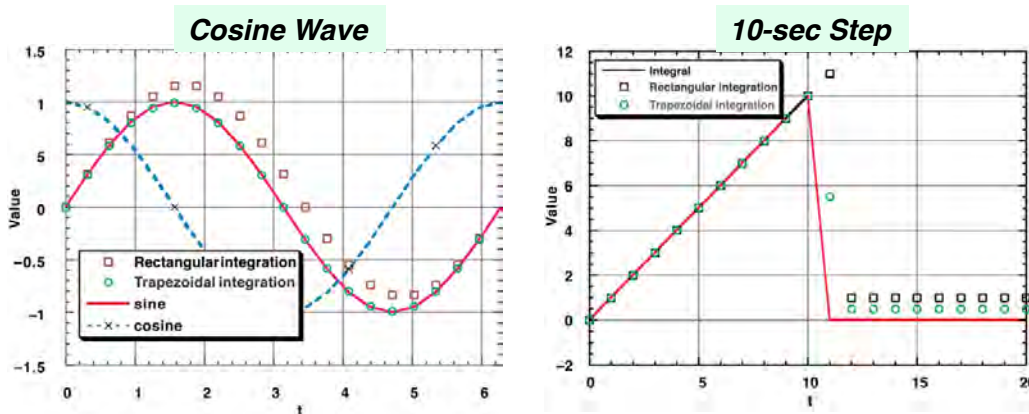
Trapezoidal (modified Euler) Integration (*ode23*)

$$\begin{aligned} \mathbf{x}(t_k) &\approx \mathbf{x}(t_{k-1}) + \frac{1}{2} [\Delta \mathbf{x}_1 + \Delta \mathbf{x}_2] \\ &\text{where} \\ \Delta \mathbf{x}_1 &= \mathbf{f}[\mathbf{x}(t_{k-1}), \mathbf{u}(t_{k-1}), \mathbf{w}(t_{k-1})] \Delta t \\ \Delta \mathbf{x}_2 &= \mathbf{f}[\mathbf{x}(t_{k-1}) + \Delta \mathbf{x}_1, \mathbf{u}(t_k), \mathbf{w}(t_k)] \Delta t \end{aligned}$$

ode23 varies step size, Δt , to reduce numerical error

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Numerical Integration Examples



How can approximation accuracy be improved?

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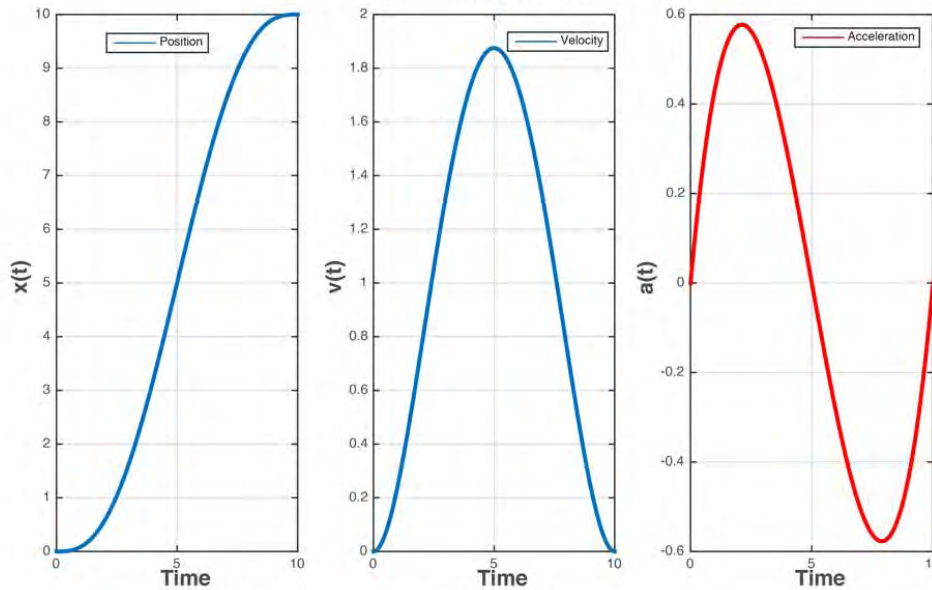
More Complicated Algorithms (e.g., MATLAB)

Solver	Problem Type	Order of Accuracy	When to Use
ode45	Nonstiff	Medium	Most of the time. This should be the first solver you try.
ode23	Nonstiff	Low	For problems with crude error tolerances or for solving moderately stiff problems.
ode113	Nonstiff	Low to high	For problems with stringent error tolerances or for solving computationally intensive problems.
ode15s	Stiff	Low to medium	If ode45 is slow because the problem is stiff.
ode23s	Stiff	Low	If using crude error tolerances to solve stiff systems and the mass matrix is constant.
ode23t	Moderately Stiff	Low	For moderately stiff problems if you need a solution without numerical damping.
ode23tb	Stiff	Low	If using crude error tolerances to solve stiff systems.

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1-D Example

$$a(0) = 0, j(0) = 0.6, s(0) = -0.36, c = 0.072$$



$$a_{net}(t) = (0) + 0.6t - 0.36t^2/2 + 0.072t^3/6$$

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Comparison of Exact and Numerically Integrated Trajectories

Calculate trajectory, given constants for $t_f = 10$

$$\begin{bmatrix} x(t) \\ v(t) \\ a(t) \end{bmatrix} = \begin{bmatrix} 1 & t & t^2/2 & t^3/6 & t^4/24 & t^5/120 \\ 0 & 1 & t & t^2/2 & t^3/6 & t^4/24 \\ 0 & 0 & 1 & t & t^2/2 & t^3/6 \end{bmatrix} \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0.6 \\ -0.36 \\ 0.072 \end{bmatrix}$$

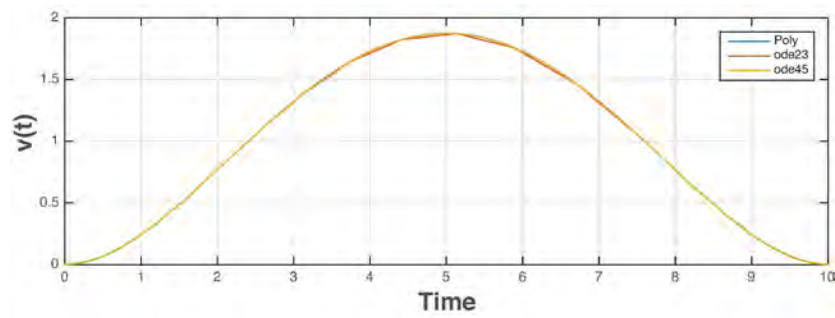
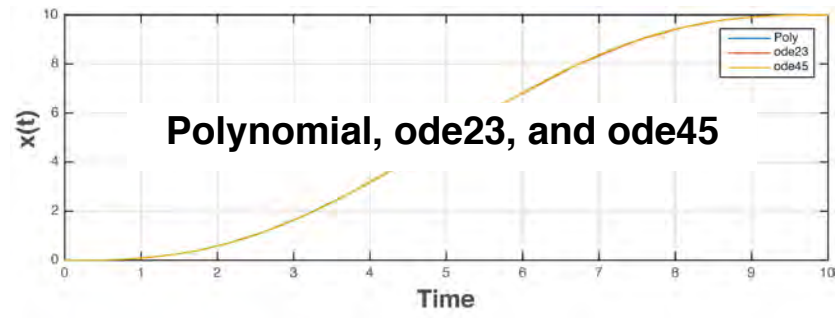
Calculate trajectory by numerical integration

$$\dot{x}_1(t) = x_2(t)$$

$$\dot{x}_2(t) = \dot{v}(t) = a(t) = 0.6t - 0.36t^2/2 + 0.072t^3/6$$

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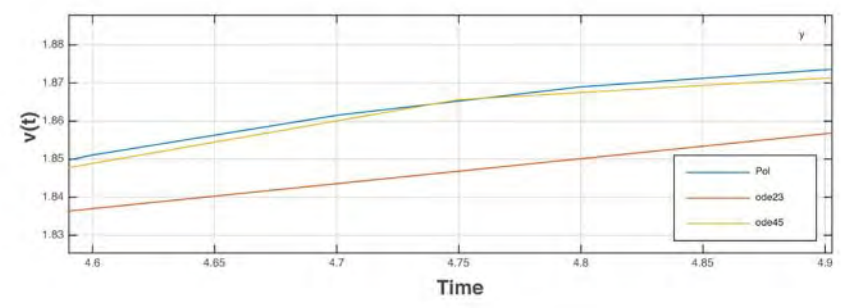
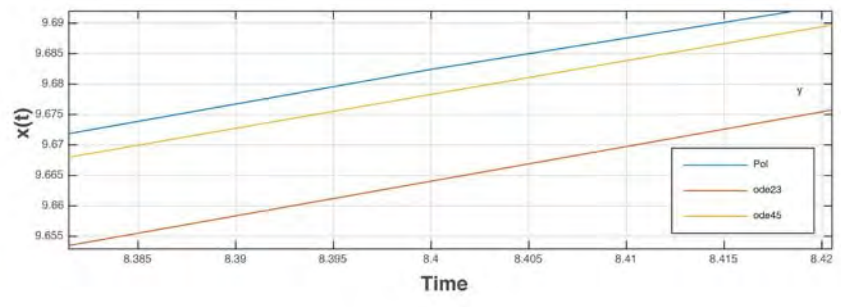
Comparison of Exact and Numerically Integrated Trajectories



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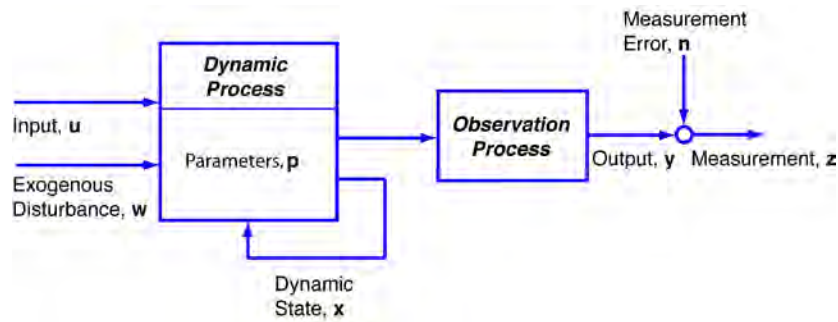
Comparison of Exact and Numerically Integrated Trajectories (Zoom)

Polynomial, ode23, and ode45



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Generic Dynamic System



Dynamic Process: Current state may depend on prior state

x : state $dim = (n \times 1)$

u : input $dim = (m \times 1)$

w : disturbance $dim = (s \times 1)$

p : parameter $dim = (\ell \times 1)$

t : time (independent variable, 1×1)

Observation Process: Measurement may contain error or be incomplete

y : output (error-free)

$dim = (r \times 1)$

n : measurement error

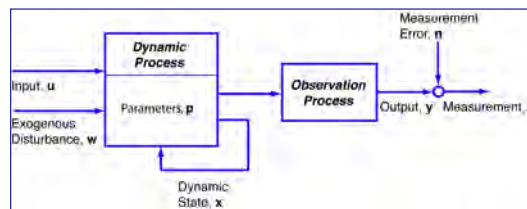
$dim = (r \times 1)$

z : measurement

$dim = (r \times 1)$

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Equations of the System



Dynamic Equation

$$\dot{\mathbf{x}}(t) = \frac{d\mathbf{x}(t)}{dt} = \mathbf{f}[\mathbf{x}(t), \mathbf{u}(t), \mathbf{w}(t), \mathbf{p}(t), t]$$

Output Equation

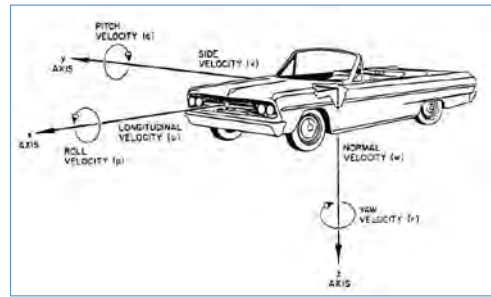
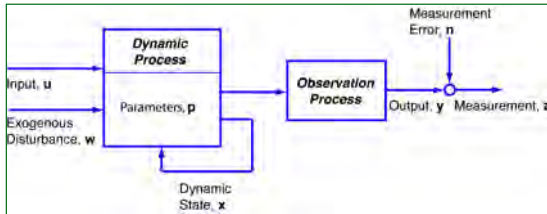
$$\mathbf{y}(t) = \mathbf{h}[\mathbf{x}(t), \mathbf{u}(t)]$$

Measurement Equation

$$\mathbf{z}(t) = \mathbf{y}(t) + \mathbf{n}(t)$$

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Dynamic System Example: Automotive Vehicle



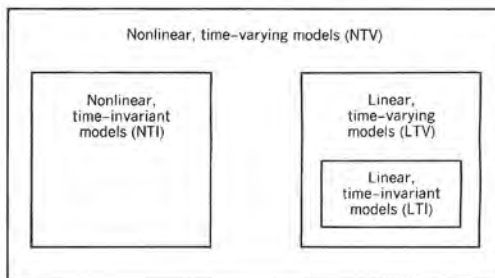
Dynamic Process

- **x** : dynamic state
 - Position, velocity, angle, angular rate
- **u** : input
 - Steering, throttle, brakes
- **w** : disturbance
 - Road surface, wind
- **p** : parameter
 - Weight, moments of inertia, drag coefficient, spring constants
- **t** : time (independent variable)

Observation Process

- **y** : error-free output
 - Speed, front-wheel angle, engine rpm, acceleration, yaw rate, throttle, brakes, GPS location
- **n** : measurement error
 - Perturbations to y
- **z** : measurement
 - Sum of y and n

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Nonlinearity and Time Variation in Dynamic Systems

Nonlinear, time-varying dynamics

$$\dot{\mathbf{x}}(t) = \mathbf{f}[\mathbf{x}(t), \mathbf{u}(t), \mathbf{w}(t), \mathbf{p}(t), t]$$

NTV

Nonlinear, time-invariant dynamics

$$\dot{\mathbf{x}}(t) = \mathbf{f}[\mathbf{x}(t), \mathbf{u}(t), \mathbf{w}(t)]$$

NTI

Linear, time-varying dynamics

$$\dot{\mathbf{x}}(t) = \mathbf{F}(t)\mathbf{x}(t) + \mathbf{G}(t)\mathbf{u}(t) + \mathbf{L}(t)\mathbf{w}(t)$$

LTV

Linear, time-invariant dynamics

$$\dot{\mathbf{x}}(t) = \mathbf{F}\mathbf{x}(t) + \mathbf{G}\mathbf{u}(t) + \mathbf{L}\mathbf{w}(t)$$

LTI

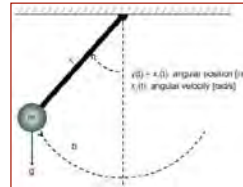
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Nonlinearity and Time Variation in Dynamic Systems

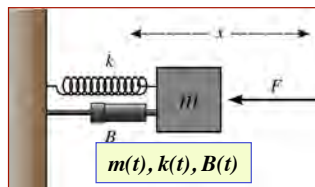
Nonlinear, time-varying dynamics



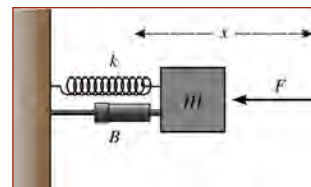
Nonlinear, time-invariant dynamics



Linear, time-varying dynamics



Linear, time-invariant dynamics



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Solutions of Ordinary Differential Dynamic Equations

Time-Domain Model (ODE)	Solution by Numerical Integration	Principle of Superposition	Frequency-Domain Model
. Nonlinear, time-varying (NTV)	Yes	No	No
. Nonlinear, time-invariant (NTI)	Yes	No	Yes (amplitude-dependent, harmonics)
. Linear, time-varying (LTV)	Yes	Yes	Approximate
. Linear, time-invariant (LTI)	Yes	Yes	Yes

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Comparison of Damped Linear and Nonlinear Systems

Linear Spring

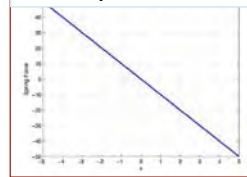
$$\dot{x}_1(t) = x_2(t)$$

$$\dot{x}_2(t) = -10x_1(t) - x_2(t)$$

Spring

Damper

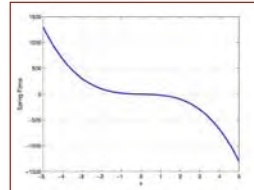
Spring Force vs. Displacement



Linear plus Stiffening Cubic Spring

$$\dot{x}_1(t) = x_2(t)$$

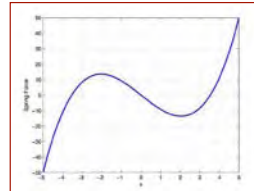
$$\dot{x}_2(t) = -10x_1(t) - 10x_1^3(t) - x_2(t)$$



Linear plus Weakening Cubic Spring

$$\dot{x}_1(t) = x_2(t)$$

$$\dot{x}_2(t) = -10x_1(t) + 0.8x_1^3(t) - x_2(t)$$



NTV or NTI?

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MATLAB Simulation of Linear and Nonlinear Dynamic Systems

MATLAB Main Program

```
% Nonlinear and Linear Examples
clear
tspan = [0 10];
xo = [0, 10];
[t1,x1] = ode23('NonLin',tspan,xo);
xo = [0, 1];
[t2,x2] = ode23('NonLin',tspan,xo);
xo = [0, 10];
[t3,x3] = ode23('Lin',tspan,xo);
xo = [0, 1];
[t4,x4] = ode23('Lin',tspan,xo);

subplot(2,1,1)
plot(t1,x1(:,1),'k',t2,x2(:,1),'b',t3,x3(:,1),'r',t4,x4(:,1),'g')
ylabel('Position'), grid
subplot(2,1,2)
plot(t1,x1(:,2),'k',t2,x2(:,2),'b',t3,x3(:,2),'r',t4,x4(:,2),'g')
xlabel('Time'), ylabel('Rate'), grid
```

Linear Spring

$$\dot{x}_1(t) = x_2(t)$$

$$\dot{x}_2(t) = -10x_1(t) - x_2(t)$$

```
function xdot = Lin(t,x)
% Linear Ordinary Differential Equation
% x(1) = Position
% x(2) = Rate
xdot = [x(2)
        -10*x(1) - x(2)];
```

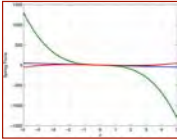
Weakening Spring

$$\dot{x}_1(t) = x_2(t)$$

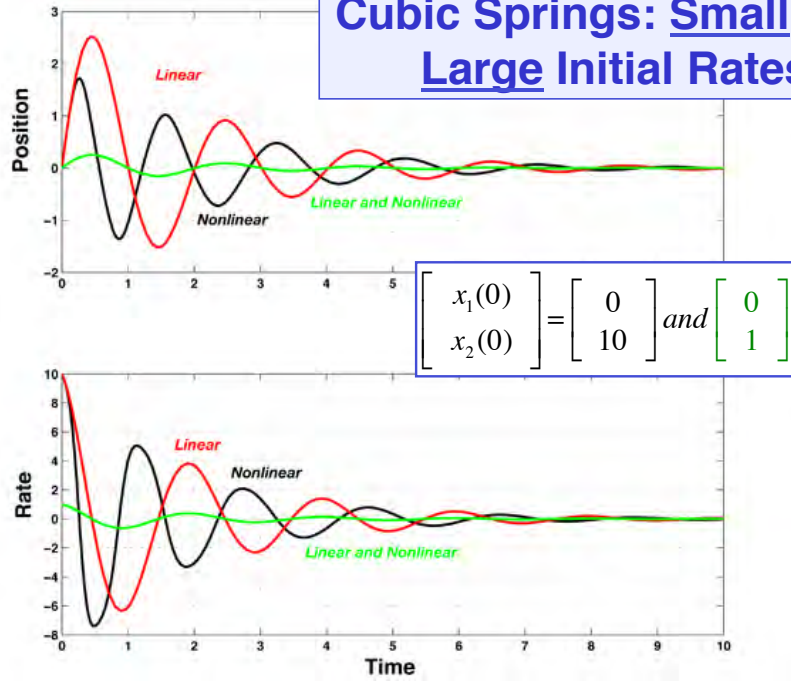
$$\dot{x}_2(t) = -10x_1(t) + 0.8x_1^3(t) - x_2(t)$$

```
function xdot = NonLin(t,x)
% Nonlinear Ordinary Differential Equation
% x(1) = Position
% x(2) = Rate
xdot = [x(2)
        -10*x(1) + 0.8*x(1)^3 - x(2)];
```

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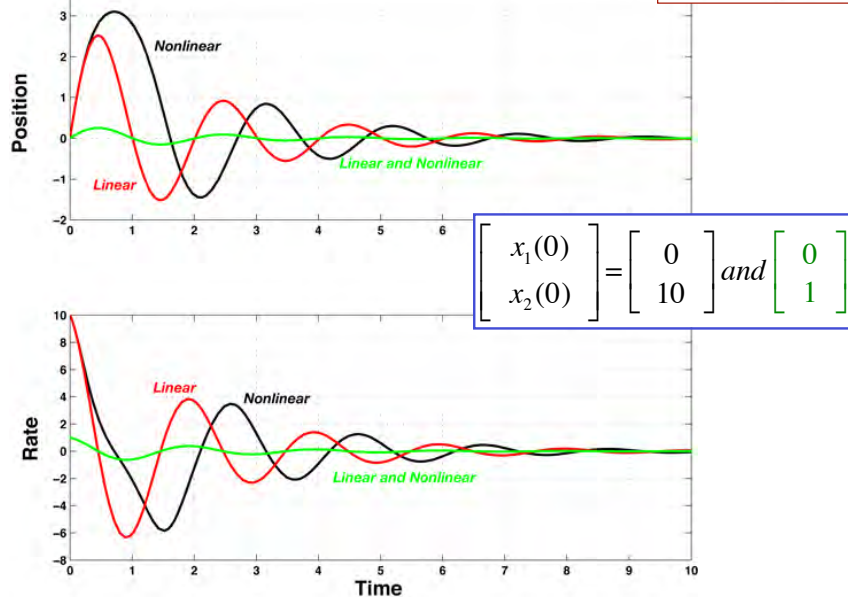
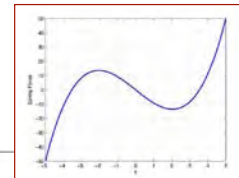


Linear and **Stiffening Cubic Springs: Small and Large Initial Rates**



Linear and nonlinear responses are indistinguishable with small initial condition

Linear and **Weakening Cubic Springs: Small and Large Initial Rates**



Linearization of Nonlinear Equations

- Given

- Nominal (or reference) robot trajectory, control, and disturbance histories

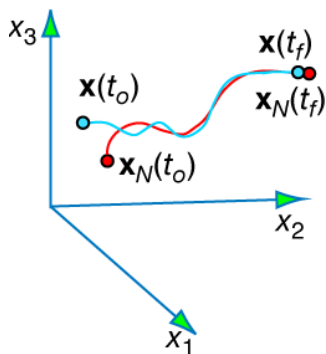
$$\mathbf{x}_N(t), \mathbf{u}_N(t), \mathbf{w}_N(t) \quad \text{for } t \text{ in } [t_o, t_f]$$

- Actual path and inputs

$$\mathbf{x}(t), \mathbf{u}(t), \mathbf{w}(t) \quad \text{for } t \text{ in } [t_o, t_f]$$

- perturbed by

- Initial condition variation
- Control variation
- Disturbance variation



$$\begin{aligned} \dim(\mathbf{x}) &= n \times 1 \\ \dim(\mathbf{u}) &= m \times 1 \\ \dim(\mathbf{w}) &= s \times 1 \end{aligned}$$

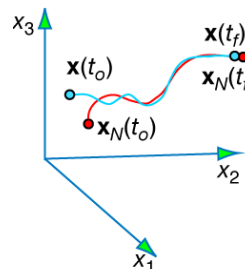
31

State, Control, and Disturbance Perturbations

Difference between nominal and actual paths:

$$\Delta \mathbf{x}(t_o) = \mathbf{x}(t_o) - \mathbf{x}_N(t_o)$$

$$\Delta \mathbf{x}(t) = \mathbf{x}(t) - \mathbf{x}_N(t)$$



Difference between nominal and actual inputs:

$$\Delta \mathbf{u}(t) = \mathbf{u}(t) - \mathbf{u}_N(t) \quad [\text{Control perturbation}]$$

$$\Delta \mathbf{w}(t) = \mathbf{w}(t) - \mathbf{w}_N(t) \quad [\text{Disturbance perturbation}]$$

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Linear Approximation of Perturbation Effects

Nominal and actual paths both satisfy same dynamic equations

$$\begin{aligned}\dot{\mathbf{x}}_N(t) &= \mathbf{f}[\mathbf{x}_N(t), \mathbf{u}_N(t), \mathbf{w}_N(t), t] \\ \dot{\mathbf{x}}(t) &= \mathbf{f}[\mathbf{x}(t), \mathbf{u}(t), \mathbf{w}(t), t]\end{aligned}$$

Actual dynamics expressed as sum of nominal terms plus perturbations

$$\begin{aligned}\dot{\mathbf{x}}(t) &= \dot{\mathbf{x}}_N(t) + \Delta\dot{\mathbf{x}}(t) \\ &= \mathbf{f}\{[\mathbf{x}_N(t) + \Delta\mathbf{x}(t)], [\mathbf{u}_N(t) + \Delta\mathbf{u}(t)], [\mathbf{w}_N(t) + \Delta\mathbf{w}(t)], t\}\end{aligned}$$

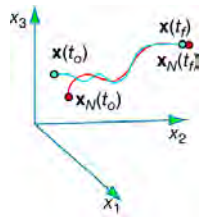
Exact

$$\approx \mathbf{f}[\mathbf{x}_N(t), \mathbf{u}_N(t), \mathbf{w}_N(t), t] + \frac{\partial \mathbf{f}}{\partial \mathbf{x}}(t) \Delta\mathbf{x}(t) + \frac{\partial \mathbf{f}}{\partial \mathbf{u}}(t) \Delta\mathbf{u}(t) + \frac{\partial \mathbf{f}}{\partial \mathbf{w}}(t) \Delta\mathbf{w}(t)$$

Approx.

Partial-derivative (*Jacobian*) matrices are evaluated along the nominal path

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Linearized Equation Approximates Perturbation Dynamics

Solve the nominal and perturbation parts *separately*
Nominal (nonlinear) equation

$$\dot{\mathbf{x}}_N(t) = \mathbf{f}[\mathbf{x}_N(t), \mathbf{u}_N(t), \mathbf{w}_N(t), t], \quad \mathbf{x}_N(0) \text{ given}$$

Perturbation (linear) equation

$$\Delta\dot{\mathbf{x}}(t) \approx \mathbf{F}(t)\Delta\mathbf{x}(t) + \mathbf{G}(t)\Delta\mathbf{u}(t) + \mathbf{L}(t)\Delta\mathbf{w}(t), \quad \Delta\mathbf{x}(0) \text{ given}$$

Approximate total solution

$$\mathbf{x}(t) \approx \mathbf{x}_N(t) + \Delta\mathbf{x}(t)$$

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Jacobian Matrices Express Sensitivity to Small Perturbations Along Nominal Trajectory

$$\mathbf{F}(t) = \frac{\partial \mathbf{f}}{\partial \mathbf{x}} \Big|_{\substack{\mathbf{x}=\mathbf{x}_N(t) \\ \mathbf{u}=\mathbf{u}_N(t) \\ \mathbf{w}=\mathbf{w}_N(t)}} ; \quad \mathbf{G}(t) = \frac{\partial \mathbf{f}}{\partial \mathbf{u}} \Big|_{\substack{\mathbf{x}=\mathbf{x}_N(t) \\ \mathbf{u}=\mathbf{u}_N(t) \\ \mathbf{w}=\mathbf{w}_N(t)}} ; \quad \mathbf{L}(t) = \frac{\partial \mathbf{f}}{\partial \mathbf{w}} \Big|_{\substack{\mathbf{x}=\mathbf{x}_N(t) \\ \mathbf{u}=\mathbf{u}_N(t) \\ \mathbf{w}=\mathbf{w}_N(t)}}$$

Sensitivity to state perturbations: **stability matrix**

$$\mathbf{F}(t) = \frac{\partial \mathbf{f}}{\partial \mathbf{x}} \Big|_{\substack{\mathbf{x}=\mathbf{x}_N(t) \\ \mathbf{u}=\mathbf{u}_N(t) \\ \mathbf{w}=\mathbf{w}_N(t)}} = \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} & \dots & \frac{\partial f_1}{\partial x_n} \\ \frac{\partial f_2}{\partial x_1} & \frac{\partial f_2}{\partial x_2} & \dots & \frac{\partial f_2}{\partial x_n} \\ \dots & \dots & \dots & \dots \\ \frac{\partial f_n}{\partial x_1} & \frac{\partial f_n}{\partial x_2} & \dots & \frac{\partial f_n}{\partial x_n} \end{bmatrix}_{\substack{\mathbf{x}=\mathbf{x}_N(t) \\ \mathbf{u}=\mathbf{u}_N(t) \\ \mathbf{w}=\mathbf{w}_N(t)}}$$

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Sensitivity to Small Control and Disturbance Perturbations Along Nominal Trajectory

$$\mathbf{F}(t) = \frac{\partial \mathbf{f}}{\partial \mathbf{x}} \Big|_{\substack{\mathbf{x}=\mathbf{x}_N(t) \\ \mathbf{u}=\mathbf{u}_N(t) \\ \mathbf{w}=\mathbf{w}_N(t)}} ; \quad \mathbf{G}(t) = \frac{\partial \mathbf{f}}{\partial \mathbf{u}} \Big|_{\substack{\mathbf{x}=\mathbf{x}_N(t) \\ \mathbf{u}=\mathbf{u}_N(t) \\ \mathbf{w}=\mathbf{w}_N(t)}} ; \quad \mathbf{L}(t) = \frac{\partial \mathbf{f}}{\partial \mathbf{w}} \Big|_{\substack{\mathbf{x}=\mathbf{x}_N(t) \\ \mathbf{u}=\mathbf{u}_N(t) \\ \mathbf{w}=\mathbf{w}_N(t)}}$$

Control-effect matrix

$$\mathbf{G}(t) = \frac{\partial \mathbf{f}}{\partial \mathbf{u}} \Big|_{\substack{\mathbf{x}=\mathbf{x}_N(t) \\ \mathbf{u}=\mathbf{u}_N(t) \\ \mathbf{w}=\mathbf{w}_N(t)}} = \begin{bmatrix} \frac{\partial f_1}{\partial u_1} & \dots & \frac{\partial f_1}{\partial u_m} \\ \dots & \dots & \dots \\ \frac{\partial f_n}{\partial u_1} & \dots & \frac{\partial f_n}{\partial u_m} \end{bmatrix}_{\substack{\mathbf{x}=\mathbf{x}_N(t) \\ \mathbf{u}=\mathbf{u}_N(t) \\ \mathbf{w}=\mathbf{w}_N(t)}}$$

Disturbance-effect matrix

$$\mathbf{L}(t) = \frac{\partial \mathbf{f}}{\partial \mathbf{w}} \Big|_{\substack{\mathbf{x}=\mathbf{x}_N(t) \\ \mathbf{u}=\mathbf{u}_N(t) \\ \mathbf{w}=\mathbf{w}_N(t)}} = \begin{bmatrix} \frac{\partial f_1}{\partial w_1} & \dots & \frac{\partial f_1}{\partial w_s} \\ \dots & \dots & \dots \\ \frac{\partial f_n}{\partial w_1} & \dots & \frac{\partial f_n}{\partial w_s} \end{bmatrix}_{\substack{\mathbf{x}=\mathbf{x}_N(t) \\ \mathbf{u}=\mathbf{u}_N(t) \\ \mathbf{w}=\mathbf{w}_N(t)}}$$

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Stiffening Cubic Spring Example

Nonlinear equation, no inputs

$$\begin{aligned} \dot{x}_1(t) &= f_1 = x_2(t) \\ \dot{x}_2(t) &= f_2 = -10x_1(t) - 10x_1^3(t) - x_2(t) \end{aligned}$$

Integrate nonlinear equation to produce nominal path

$$\begin{bmatrix} x_1(0) \\ x_2(0) \end{bmatrix} \rightarrow \int_0^{t_f} \begin{bmatrix} f_{1_N} \\ f_{2_N} \end{bmatrix} dt \rightarrow \begin{bmatrix} x_{1_N}(t) \\ x_{2_N}(t) \end{bmatrix} \text{ in } [0, t_f]$$

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Stiffening Cubic Spring Example

Evaluate partial derivatives along the path

$$\mathbf{F}(t) = \begin{bmatrix} \frac{\partial f_1}{\partial x_1} = 0 & \frac{\partial f_1}{\partial x_2} = 1 \\ \frac{\partial f_2}{\partial x_1} = -10 - 30x_{1_N}^2(t) & \frac{\partial f_2}{\partial x_2} = -1 \end{bmatrix}$$

$$\mathbf{G}(t) = \begin{bmatrix} \frac{\partial f_1}{\partial u} = 0 \\ \frac{\partial f_2}{\partial u} = 0 \end{bmatrix}$$

$$\mathbf{L}(t) = \begin{bmatrix} \frac{\partial f_1}{\partial w} = 0 \\ \frac{\partial f_2}{\partial w} = 0 \end{bmatrix}$$

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Nominal and Perturbation Dynamic Equations

Nominal system, NTI

$$\begin{aligned}\dot{\mathbf{x}}_N(t) &= \mathbf{f}[\mathbf{x}_N(t)], \quad \mathbf{x}_N(0) \text{ given} \\ \dot{x}_{1_N}(t) &= x_{2_N}(t) \\ \dot{x}_{2_N}(t) &= -10x_{1_N}(t) - 10x_{1_N}^3(t) - x_{2_N}(t)\end{aligned}$$

$$\begin{bmatrix} x_{1_N}(0) \\ x_{2_N}(0) \end{bmatrix} = \begin{bmatrix} 0 \\ 9 \end{bmatrix}$$

New initial condition:

Perturbation system, LTV

$$\Delta \dot{\mathbf{x}}(t) = \mathbf{F}(t)\Delta \mathbf{x}(t), \quad \Delta \mathbf{x}(0) \text{ given}$$

$$\begin{bmatrix} \Delta \dot{x}_1(t) \\ \Delta \dot{x}_2(t) \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ -(10 + 30x_{1_N}^2(t)) & -1 \end{bmatrix} \begin{bmatrix} \Delta x_1(t) \\ \Delta x_2(t) \end{bmatrix}$$

$$\begin{bmatrix} \Delta x_1(0) \\ \Delta x_2(0) \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

$x_{1_N}(t)$ generated by nominal solution

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Nominal and Perturbation Dynamic Solutions for Cubic Spring Example with $\mathbf{x}_N(0) = \mathbf{0}$

If nominal solution remains at equilibrium

$$\dot{\mathbf{x}}_N(t) = \mathbf{f}[\mathbf{x}_N(t)], \quad \mathbf{x}_N(0) = \mathbf{0}, \quad \mathbf{x}_N(t) = \mathbf{0} \text{ in } [0, \infty]$$

Linearization is time-invariant

$$\begin{bmatrix} \Delta \dot{x}_1(t) \\ \Delta \dot{x}_2(t) \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ -10 & -1 \end{bmatrix} \begin{bmatrix} \Delta x_1(t) \\ \Delta x_2(t) \end{bmatrix}$$

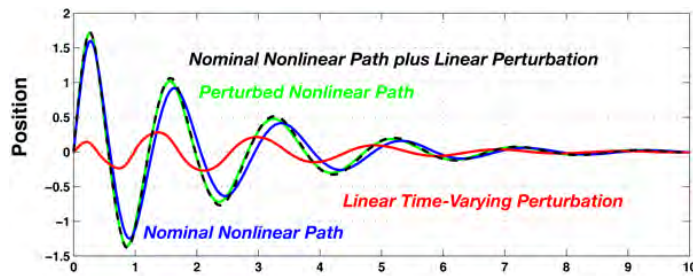
= Linear, **Time-Invariant** (LTI) System

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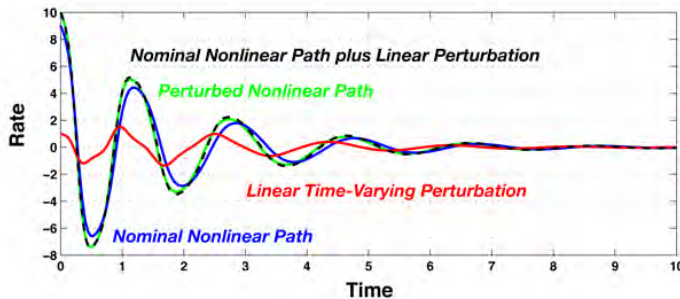
Comparison of Approximate and Exact Solutions

Rate Initial Condition:

$$\begin{aligned} x_{2_N}(0) &= 9 \\ \Delta x_2(0) &= 1 \\ x_2(t) &= 10 \\ x_{2_N}(t) + \Delta x_2(t) &= 10 \end{aligned}$$



$\mathbf{x}_N(t)$: Nominal
 $\Delta \mathbf{x}(t)$: Perturbation
 $\mathbf{x}(t)$: Actual
 $\mathbf{x}_N(t) + \Delta \mathbf{x}(t)$: Approximation



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Initial-Condition Response of a Linear, Time-Invariant (LTI) Model

- Doubling the initial condition doubles the output
- Stability, speed of response, and damping are independent of the initial condition

```
% Linear Model - Initial Condition
F = [-0.5572 -0.7814; 0.7814 0];
G = [1 -1; 0 2];
Hx = [1 0; 0 1];
sys = ss(F, G, Hx, 0);
```

```
xo = [1; 0];
[y1,t1,x1] = initial(sys, xo);
xo = [2; 0];
[y2,t2,x2] = initial(sys, xo);
```

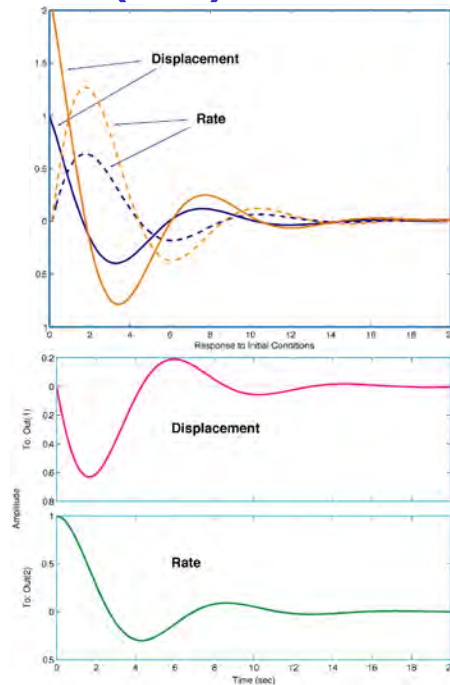
```
plot(t1,y1,t2,y2)
```

```
figure
xo = [0; 1];
initial(sys, xo)
```

$$\mathbf{F} = \begin{bmatrix} -0.5572 & -0.7814 \\ 0.7814 & 0 \end{bmatrix}$$

$$\mathbf{G} = \begin{bmatrix} 1 & -1 \\ 0 & 2 \end{bmatrix}$$

$$\mathbf{H}_x = \mathbf{I}_2; \quad \mathbf{H}_u = \mathbf{0}$$



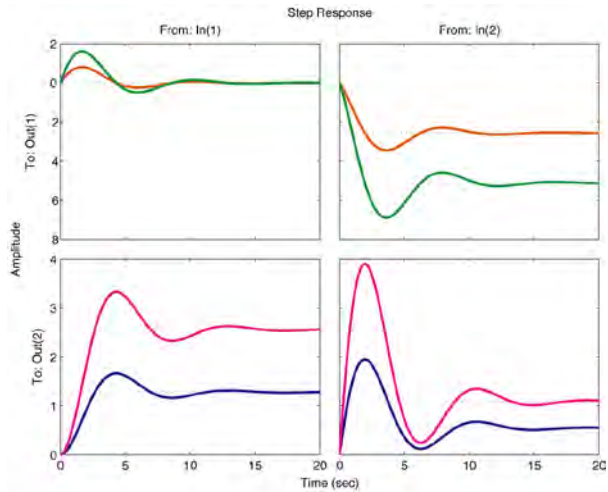
42

Step Response of a Linear, Time-Invariant Model

- **Doubling the step input doubles the output**
- **Stability, speed of response, and damping are independent of the input**

```
% Linear Model - Step
F = [-0.5572 -0.7814;0.7814 0];
G = [1 -1;0 2];
Hx = [1 0;0 1];
sys = ss(F, G, Hx,0);
sys2 = ss(F, 2*G, Hx,0);

% Step response
step(sys, sys2)
```



Response to Combined Initial Condition and Step Input

Linear system responses are additive

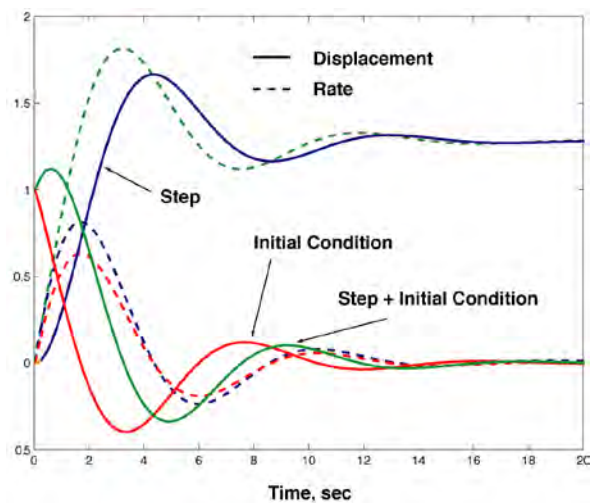
```
% Linear Model - Superposition
F = [-0.5572 -0.7814;0.7814 0];
G = [1;0];
Hx = [1 0;0 1];
sys = ss(F, G, Hx,0);

xo = [1; 0];
t = [0:0.2:20];
u = ones(1,length(t));

[y1,t1,x1] = lsim(sys,u,t,xo);
[y2,t2,x2] = lsim(sys,u,t);

u = zeros(1,length(t));
[y3,t3,x3] = lsim(sys,u,t,xo);

plot(t1,y1,t2,y2,t3,y3)
```



Initial-Condition Responses of 1st-Order LTI Systems are Exponentials

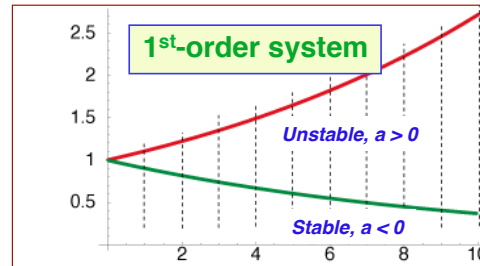
$$\Delta \dot{\mathbf{x}}(t) = \mathbf{F} \Delta \mathbf{x}(t), \quad \Delta \mathbf{x}(0) \text{ given}$$

State vector is a scalar

$$\mathbf{F} = [a]$$

$$\Delta \dot{x}(t) = a \Delta x(t)$$

$$\Delta x(0) \text{ given}$$



$$\Delta x(t) = \int_0^t \Delta \dot{x}(t) dt = \int_0^t a \Delta x(t) dt$$

$$= e^{at} \Delta x(0)$$

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Initial-Condition Responses of 2nd-Order LTI Systems Exponentials and Sinusoids

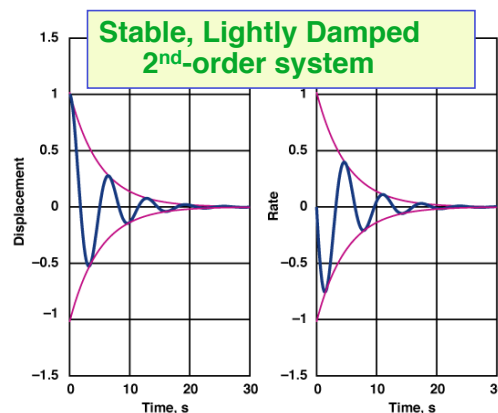
$$\Delta \dot{\mathbf{x}}(t) = \mathbf{F} \Delta \mathbf{x}(t)$$

$\Delta \mathbf{x}(0)$ given

$$\mathbf{F} = \begin{bmatrix} f_{11} & f_{12} \\ f_{21} & f_{22} \end{bmatrix}$$

$$\Delta \dot{x}_1 = f_{11} \Delta x_1 + f_{12} \Delta x_2 + g_1 \Delta u$$

$$\Delta \dot{x}_2 = f_{21} \Delta x_1 + f_{22} \Delta x_2 + g_2 \Delta u$$

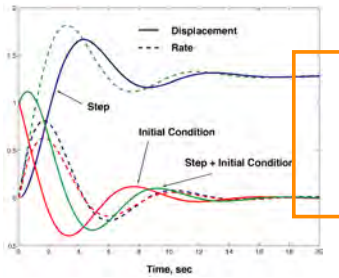


Sinusoid with exponential envelope

$$\Delta x_1(t) = A_1 e^{-\zeta \omega_n t} \cos \left[\omega_n \sqrt{1 - \zeta^2} t + \phi_1 \right]$$

$$\Delta x_2(t) = A_2 e^{-\zeta \omega_n t} \cos \left[\omega_n \sqrt{1 - \zeta^2} t + \phi_2 \right]$$

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Equilibrium Response of Linear, Time-Invariant Models

- **General equation**

$$\Delta \dot{\mathbf{x}}(t) = \mathbf{F}\Delta \mathbf{x}(t) + \mathbf{G}\Delta \mathbf{u}(t) + \mathbf{L}\Delta \mathbf{w}(t)$$
- **At equilibrium,**
 - Derivative goes to zero
 - State is unchanging
$$\mathbf{0} = \mathbf{F}\Delta \mathbf{x}(t) + \mathbf{G}\Delta \mathbf{u}(t) + \mathbf{L}\Delta \mathbf{w}(t)$$
- **State at equilibrium**

$$\begin{aligned} \Delta \mathbf{x}^* &= -\mathbf{F}^{-1}(\mathbf{G}\Delta \mathbf{u}^* + \mathbf{L}\Delta \mathbf{w}^*) \\ &= -\frac{\text{Adj}(\mathbf{F})}{\det(\mathbf{F})}(\mathbf{G}\Delta \mathbf{u}^* + \mathbf{L}\Delta \mathbf{w}^*) \end{aligned}$$

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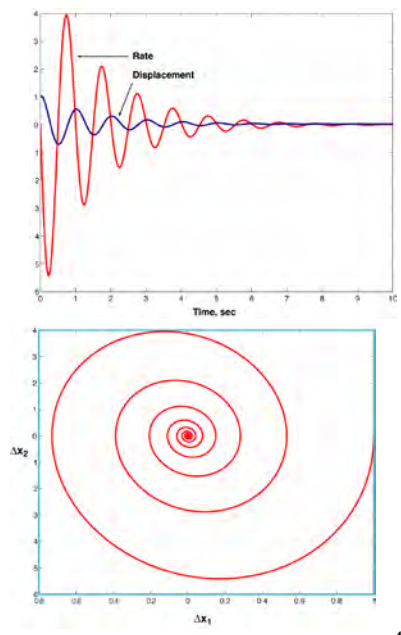
State (“Phase”)-Plane Plots

- **Cross-plot of one component against another**
- **Time or frequency not shown explicitly in phase plane**

```
% 2nd-Order Model - Initial Condition Response
clear
z = 0.1; % Damping ratio
wn = 6.28; % Natural frequency, rad/s
F = [0 1; -wn^2 -2*z*wn];
G = [1 -1; 0 2];
Hx = [1 0; 0 1];
sys = ss(F, G, Hx, 0);
t = [0:0.01:10];
xo = [1; 0];
[y1,t1,x1] = initial(sys, xo, t);

plot(t1,y1)
grid on

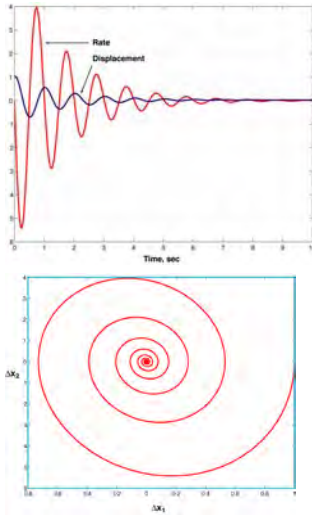
figure
plot(y1(:,1),y1(:,2))
grid on
```



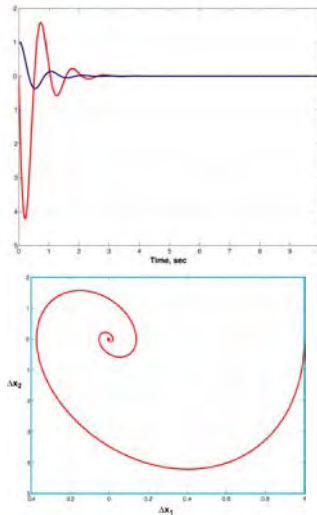
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Effects of Damping Ratio on State-Plane Plots

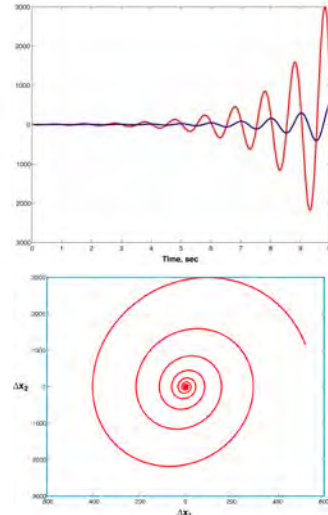
- Damping ratio = 0.1



- Damping ratio = 0.3

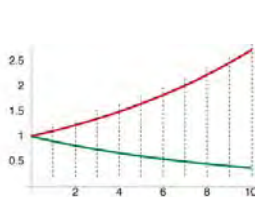


- Damping ratio = -0.1

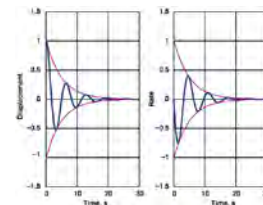


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Response of Higher-Order LTI Systems *Superposition of Sub-System Responses*



$$\mathbf{F}_{System} = \begin{bmatrix} \mathbf{F}_{System 1} & \text{Effect of \#2 on \#1} \\ \text{Effect of \#1 on \#2} & \mathbf{F}_{System 2} \end{bmatrix}$$



- Third-order system with uncoupled 1st- and 2nd-order sub-systems**

$$\mathbf{F} = \begin{bmatrix} f_{11} & 0 & 0 \\ 0 & 0 & 1 \\ 0 & f_{32} & f_{33} \end{bmatrix}$$

$$\Delta \mathbf{x} = \begin{bmatrix} \Delta x_1 \\ \Delta x_2 \\ \Delta x_3 \end{bmatrix}$$

- Coupling in first row and first column**

$$\mathbf{F} = \begin{bmatrix} f_{11} & f_{12} & f_{13} \\ f_{21} & 0 & 1 \\ f_{31} & f_{32} & f_{33} \end{bmatrix}$$

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Examples of Coupled and Uncoupled Third-Order Systems

Third-order system with uncoupled 1st- and 2nd-order sub-systems

$$\mathbf{F} = \begin{bmatrix} -1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & -1 & -1.414 \end{bmatrix}$$

$$\Delta \mathbf{x} = \begin{bmatrix} \Delta x_1 \\ \Delta x_2 \\ \Delta x_3 \end{bmatrix}$$

Coupling in first row and first column

Position Coupling, Δx_3

$$\mathbf{F} = \begin{bmatrix} -1 & 0.1 & 0 \\ 0.1 & 0 & 1 \\ 0 & -1 & -1.414 \end{bmatrix}$$

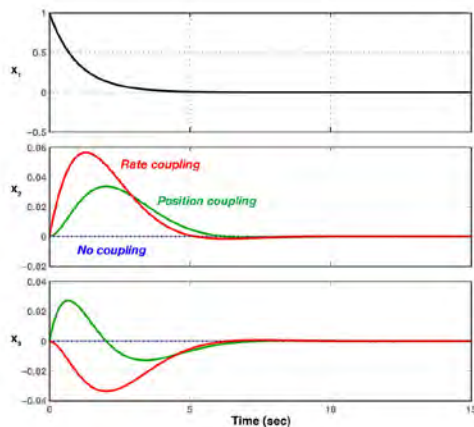
Rate Coupling, Δx_2

$$\mathbf{F} = \begin{bmatrix} -1 & 0 & 0.1 \\ 0 & 0 & 1 \\ 0.1 & -1 & -1.414 \end{bmatrix}$$

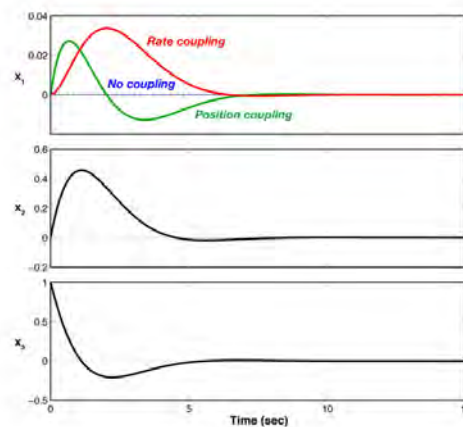
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3rd-Order LTI Systems with Coupled Response

Initial Condition on Δx_1



Initial Condition on Δx_3



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*Next Time:
Dynamic Effects of
Feedback Control*

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Supplemental Material

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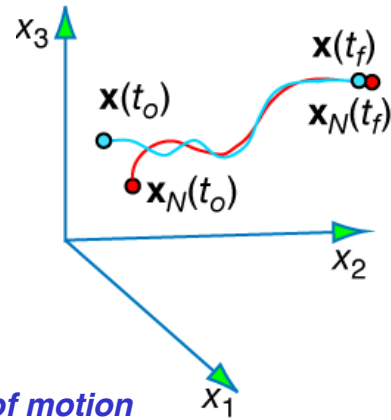
Perturbed Initial Conditions Produce Perturbed Path

- Given**

- Initial condition, control, and disturbance histories

$$\mathbf{x}(t_0), \mathbf{u}(t), \mathbf{w}(t) \quad \text{for } t \text{ in } [t_0, t_f]$$

- Path (or **trajectory**) is approximated by executing a numerical algorithm
- **Perturbing the initial condition produces a new path**



- Both paths satisfy the same equations of motion**

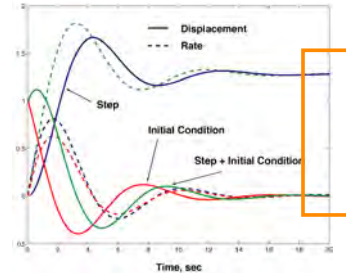
$$\dot{\mathbf{x}}_N(t) = \mathbf{f}[\mathbf{x}_N(t), \mathbf{u}(t), \mathbf{w}(t), t], \quad \mathbf{x}_N(t_0) \text{ given}$$

$$\dot{\mathbf{x}}(t) = \mathbf{f}[\mathbf{x}(t), \mathbf{u}(t), \mathbf{w}(t), t], \quad \mathbf{x}(t_0) \text{ given}$$

- \mathbf{x}_N : Nominal path
- \mathbf{x} : Perturbed path

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Equilibrium Response of Second-Order LTI System



System description

$$\mathbf{F} = \begin{bmatrix} f_{11} & f_{12} \\ f_{21} & f_{22} \end{bmatrix}; \quad \mathbf{G} = \begin{bmatrix} g_1 \\ g_2 \end{bmatrix}; \quad \mathbf{L} = \begin{bmatrix} l_1 \\ l_2 \end{bmatrix}$$

State equilibrium depends on constant input values

$$\begin{bmatrix} \Delta x_1^* \\ \Delta x_2^* \end{bmatrix} = - \begin{bmatrix} f_{22} & -f_{12} \\ -f_{21} & f_{11} \end{bmatrix} \left[\begin{pmatrix} g_1 \\ g_2 \end{pmatrix} \Delta u^* + \begin{pmatrix} l_1 \\ l_2 \end{pmatrix} \Delta w^* \right]$$

$$|\mathbf{F}| = (f_{11}f_{22} - f_{12}f_{21}) \neq 0$$

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