Ch3: Modeling of Dynamic Systems – Part 2

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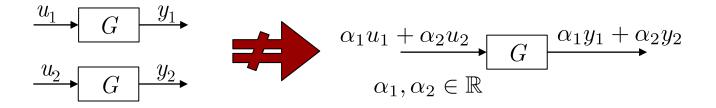


Nonlinear Systems



Nonlinear Systems

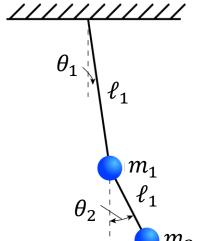
A system is nonlinear if the **principle of superposition** does **not** apply.



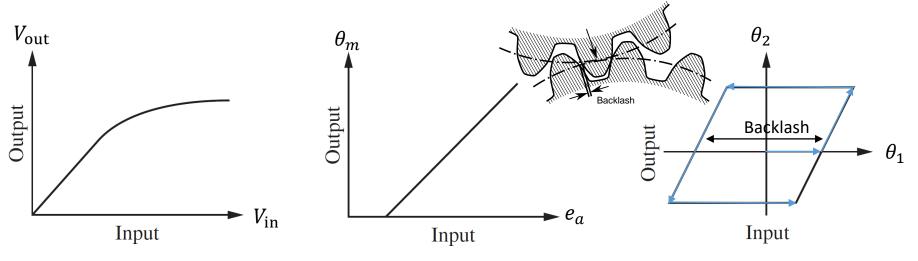
For example, in the **dynamic equations of robots** usually the nonlinear terms sin , cos , and squares of velocities appears.

Double-Pendulum:

$$(m_1 + m_2)\ell_1\ddot{\theta}_1 + m_2\ell_2\ddot{\theta}_2\cos(\theta_1 - \theta_2) + m_2\ell_2\dot{\theta}_2^2\sin(\theta_1 - \theta_2) + g(m_1 + m_2)\sin\theta_1 = 0$$
$$m_2\ell_2\ddot{\theta}_2 + m_2\ell_1\ddot{\theta}_1\cos(\theta_1 - \theta_2) - m_2\ell_1\dot{\theta}_1^2\sin(\theta_1 - \theta_2) + m_2g\sin\theta_2 = 0$$



Examples of Physical Nonlinearities



Amplifier Saturation

An electronic amplifier is linear over a specific range but exhibits the nonlinearity called saturation at high input voltages.

Motor Dead Zone

A motor that does not respond at very low input voltages due to frictional forces exhibits a nonlinearity called dead zone.

Backlash in Gears

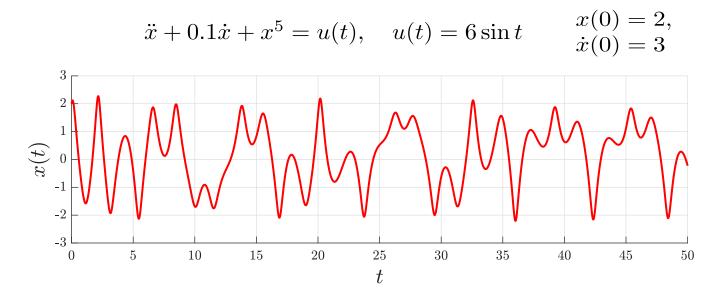
Gears that do not fit tightly exhibit a nonlinearity called backlash which the input moves over a small range without the output responding.

 Nonlinearities can be classified in terms of their mathematical properties, as continuous and discontinuous. Because discontinuous nonlinearities cannot be locally approximated by linear functions, they are also called hard nonlinearities (e.g., backlash, hysteresis, or stiction).



Nonlinear System Behavior: Chaos

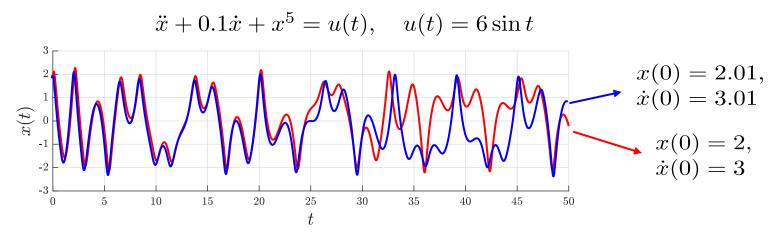
• In the steady state, **sinusoidal inputs** to a stable LTI system generate a sinusoidal outputs of the same frequency (but different in amplitude and phase angle from the input). By contrast, the output of a nonlinear system may display sinusoidal, periodic, or chaotic behaviors.



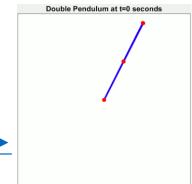


Nonlinear System Behavior: Chaos

 For stable linear systems, small differences in initial conditions can only cause small differences in output. However, output of strongly nonlinear systems is extremely sensitive to initial conditions.



• Starting the pendulum from a slightly different initial condition would result in a vastly different trajectory.



Nonlinear Systems	Linearization of Nonlinear Systems	State-Space Representation	MATLAB
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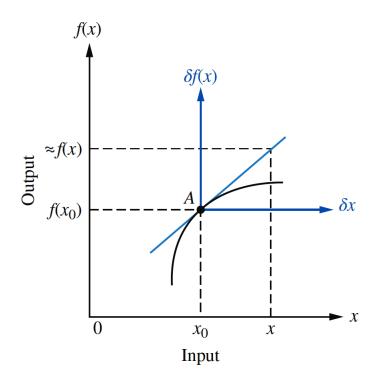
Linearization of Nonlinear Systems



Linearization of Nonlinear Systems

In control engineering, a normal operation of the system may be around an **equilibrium point** or **a limited operating range**. Therefor, it is possible to approximate the nonlinear system by an equivalent linear system within the limited operating range.

• Linear approximations simplify the analysis and design of a system.

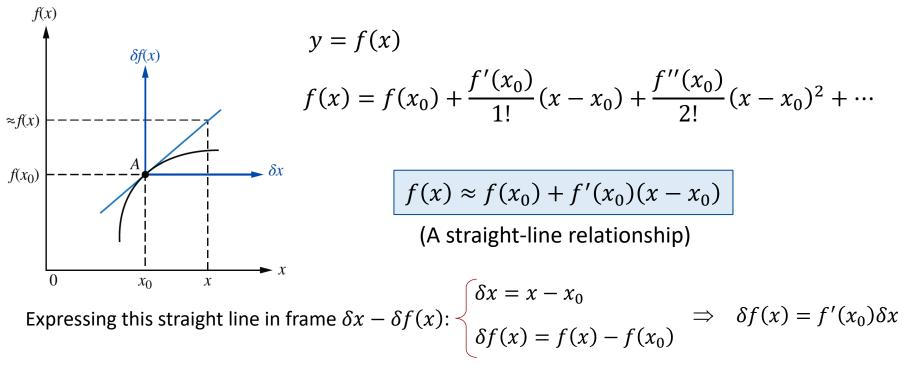




Linear Approximation of Nonlinear Mathematical Models

The linearization procedure is based on (1) the expansion of nonlinear function f(x) into a **Taylor Series** about the operating point $A(x_0, y_0 = f(x_0))$ and (2) the retention of only the linear term.

Note: Since the variables deviate only slightly from the operating condition $(x - x_0)$, higher-order terms of the Taylor series expansion can be neglected.



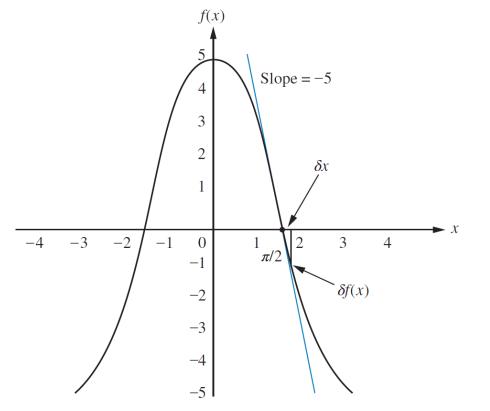
Linearize $f(x) = 5 \cos x$ about $x = \pi/2$.

Method 1:

 $f(x) \approx f(x_0) + f'(x_0)(x - x_0)$ $f(x) \approx 5\cos\left(\frac{\pi}{2}\right) - 5\sin\left(\frac{\pi}{2}\right)\left(x - \frac{\pi}{2}\right)$ $f(x) \approx -5\left(x - \frac{\pi}{2}\right)$

Method 2:

For small deviation about $\frac{\pi}{2}$: $x = \frac{\pi}{2} + \delta x$ $f(x) = 5 \cos\left(\delta x + \frac{\pi}{2}\right) = -5 \sin \delta x$ $f(x) \approx -5\delta x$





Linearize $\ddot{x} + 2\dot{x} + \cos x = 0$ for small deviations about $x = \pi/4$.

Answer:

$$\ddot{x} + 2\dot{x} - \frac{\sqrt{2}}{2}x = -\frac{\sqrt{2}}{2} - \frac{\sqrt{2}}{2}\frac{\pi}{4}$$

Nonlinear Systems	Linearization of Nonlinear Systems	State-Space Representation	MATLAB
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State-Space Representation

Amin Fakhari, Fall 2023



Some Definitions

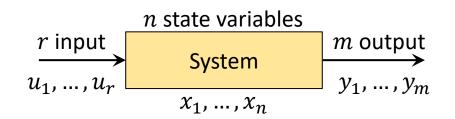
Linear Combination: A linear combination of *n* variables, x_i , is given by

 $S = k_1 x_1 + k_2 x_2 + \dots + k_n x_n$, $k_i = \text{constant}, i = 1, \dots, n$

Linear Independence: A set of variables is said to be linearly independent if none of the variables can be written as a linear combination of the others.

System Variable: Any variable that responds to an input or initial conditions in a system.

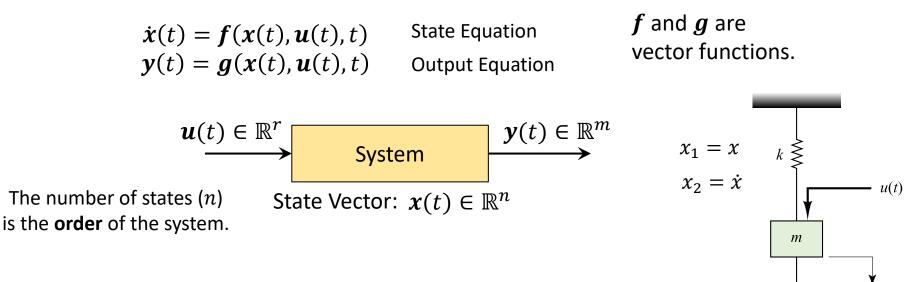
State Variables: The **smallest set of linearly independent** system variables $(x_1, ..., x_n)$ such that knowledge of these variables at $t = t_0$, together with knowledge of the input (u(t)) for $t \ge t_0$, completely determines the behavior of the system for any time $t \ge t_0$.





State-Space Representation

State-space Representation is a mathematical model of a physical system as a set of input $u(t) \in \mathbb{R}^r$, output $y(t) \in \mathbb{R}^m$, and state variables $x(t) \in \mathbb{R}^n$ related by n simultaneous first-order differential equations.



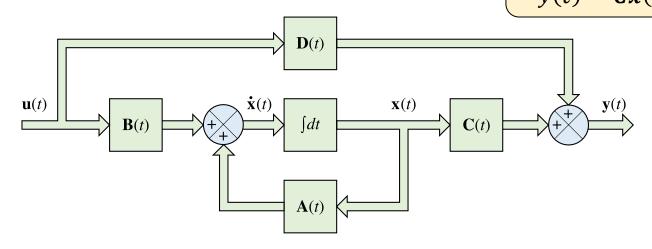
Note: State variables need not be physically measurable or observable quantities. Note: The choice of state variables of a system is not unique, but the number of states is unique. For all invertible $T \in \mathbb{R}^{n \times n}$, $\overline{x}(t) = \mathbf{T}x(t)$ can be also the system state variables.

x(t)

h

State-Space Representation

General Form: MIMO, Nonlinear, Time MIMO, Linear, Time MIMO, Linear, Variant (General Form) **Time Variant** Invariant $\dot{\boldsymbol{x}}(t) = \boldsymbol{A}(t)\boldsymbol{x}(t) + \boldsymbol{B}(t)\boldsymbol{u}(t)$ $\dot{\boldsymbol{x}}(t) = \boldsymbol{A}\boldsymbol{x}(t) + \boldsymbol{B}\boldsymbol{u}(t)$ $\dot{\boldsymbol{x}}(t) = \boldsymbol{f}(\boldsymbol{x}, \boldsymbol{u}, t)$ $\mathbf{y}(t) = \mathbf{g}(\mathbf{x}, \mathbf{u}, t)$ $\mathbf{y}(t) = \mathbf{C}(t)\mathbf{x}(t) + \mathbf{D}(t)\mathbf{u}(t)$ $\mathbf{y}(t) = \mathbf{C}\mathbf{x}(t) + \mathbf{D}\mathbf{u}(t)$ SISO, Linear, Time Invariant A: State matrix, **B**: Input matrix $\dot{\boldsymbol{x}}(t) = \boldsymbol{A}\boldsymbol{x}(t) + \boldsymbol{B}\boldsymbol{u}(t)$ **D**: Feedforward matrix *C*: Output matrix, y(t) = Cx(t) + Du(t)





State-Space Representation of LTI Systems

Consider a general, *n*th-order, linear differential equation with constant coefficients:

$$\frac{d^n x}{dt^n} + a_{n-1} \frac{d^{n-1} x}{dt^{n-1}} + \dots + a_0 x = b_0 u$$

- An *n*th-order differential equation can be converted to *n* simultaneous **first-order differential equations**.
- There are many ways to do this conversion and obtain state-space representations of systems, such as **phase-variable** form, **controllable** canonical form, **observable** canonical form, **diagonal** canonical form, and **Jordan** canonical form.

A convenient way to choose state variables is to choose x(t) and its (n - 1) derivatives as the state variables, which are called **phase variables**.



Nonlinear SystemsLinearization of Nonlinear SystemsState-Space RepresentationMATLABOOOOOO∇OOOO●∇∇0∇0∇0OO

State-Space Representation of LTI Systems

$$\frac{d^{n}x}{dt^{n}} + a_{n-1}\frac{d^{n-1}x}{dt^{n-1}} + \dots + a_{0}x = b_{0}u$$

$$x_{1} = x$$

$$x_{2} = \frac{dx}{dt}$$

$$\vdots$$

$$x_{n} = \frac{d^{n-1}x}{dt^{n-1}}$$

$$\frac{d^{n-1}x}{dt^{n-1}}$$

$$\frac{y_{2} = \frac{d^{2}x}{dt^{2}}}{dt^{2}}$$

$$\frac{y_{2} = \frac{d^{2}x}{dt^{2}}}{dt^{2}}$$

$$\frac{y_{2} = \frac{d^{2}x}{dt^{2}}}{dt^{2}}$$

$$\frac{y_{2} = x_{3}}{dt^{n-1}x}$$

$$\frac{y_{n-1} = x_{n}}{x_{n} = -a_{n-1}x_{n} - \dots - a_{0}x_{1} + b_{0}u}$$

$$\frac{y_{n-1} = x_{n}}{x_{n} = -a_{n-1}x_{n} - \dots - a_{0}x_{1} + b_{0}u}$$

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$$\frac{y_{n-1} = x_{n}}{x_{n} - u_{n-1}x_{n} - \dots - u_{0}x_{1} + b_{0}u}$$

$$\frac{y_{n-1} = u_{n-1}x_{n}}{y_{n-1} - u_{n-1}x_{n} - \dots - u_{n-1}u}$$

$$\frac{y_{n-1} = u_{n-1}x_{n}}{y_{n-1} - u_{n-1}x_{n} - \dots - u_{n-1}u}$$

MEC411 • Ch3: Modeling of Dynamic Systems – Part 2

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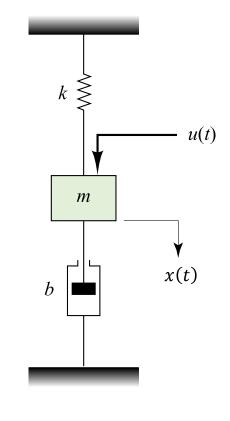
Stony Brook University



The external force u(t) is the input to the system, and the displacement x(t) of the mass, measured from the equilibrium position in the absence of the external force, is the output. Find the state equations.

Solution:

 $m\ddot{x} + b\dot{x} + kx = u$ Let's define: $\begin{array}{c} x_1 = x \\ x_2 = \dot{x} \end{array}$ $\dot{x}_1 = x_2 \\ \dot{x}_2 = \frac{1}{m}(-kx - b\dot{x}) + \frac{1}{m}u \quad \checkmark \quad \dot{x}_1 = x_2 \\ \dot{x}_2 = -\frac{k}{m}x_1 - \frac{b}{m}x_2 + \frac{1}{m}u \\ y = x = x_1 \end{array}$ $\left[\begin{array}{c} \dot{x}_1 \\ \dot{x}_2 \end{array}\right] = \begin{bmatrix} 0 & 1 \\ -\frac{k}{m} & -\frac{b}{m} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} 0 \\ 1 \\ -\frac{1}{m} \end{bmatrix} u \qquad y = \begin{bmatrix} 1 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$



 $[\]dot{x} = Ax + Bu$ y = Cx



Find the state equations. What is the output equation if the output is $z_1(t)$?

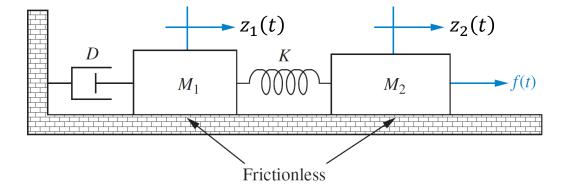
Solution:

$$M_1 \ddot{z}_1 + D \dot{z}_1 + K z_1 - K z_2 = 0$$

$$M_2 \ddot{z}_2 - K z_1 + K z_2 = f(t)$$

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Converting from SS to a TF

Deriving the transfer function from the state-space equations:

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

$$\overset{\text{Laplace transform}}{\text{initial conditions}} sX(s) = AX(s) + BU(s) \rightarrow X(s) = (sI - A)^{-1}BU(s)$$

$$y = Cx + Du$$

$$Y(s) = CX(s) + DU(s)$$

$$(I \text{ is the identity matrix})$$

$$Y(s) = [C(sI - A)^{-1}B + D]U(s)$$
Transfer Function Matrix

Transfer Function for a SISO system which U(s) = U(s) and Y(s) = Y(s):

$$G(s) = \frac{Y(s)}{U(s)} = \boldsymbol{C}(s\boldsymbol{I} - \boldsymbol{A})^{-1}\boldsymbol{B} + D$$



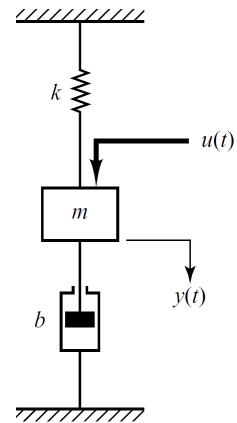
Obtain the transfer function Y(s)/U(s) from the state-space equations of the system shown in the previous example.

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

Solution:

$$G(s) = \frac{Y(s)}{U(s)} = \boldsymbol{C}(s\boldsymbol{I} - \boldsymbol{A})^{-1}\boldsymbol{B} + D$$

$$G(s) = \frac{1}{ms^2 + bs + k}$$





Converting a TF to SS

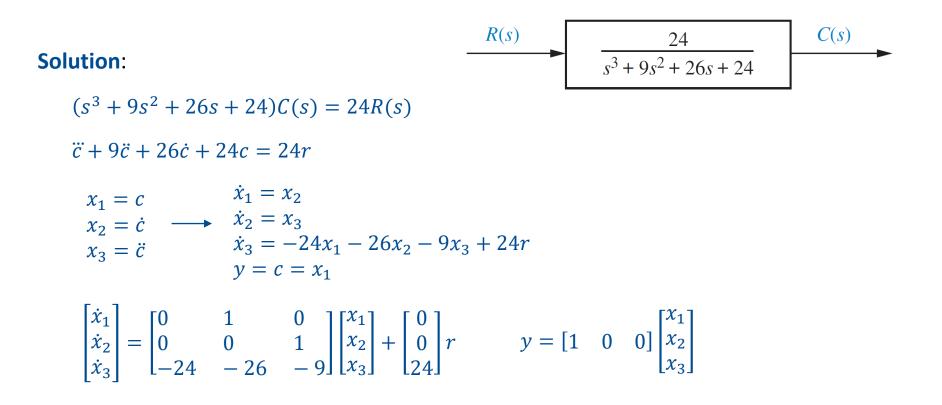
To convert a transfer function into state-space equations in phase-variable form, first convert the transfer function to a **differential equation** by cross-multiplying and taking the inverse Laplace transform, assuming zero initial conditions.

$$\frac{Y(s)}{U(s)} = \frac{b_0}{s^n + a_{n-1}s^{n-1} + \dots + a_0} \longrightarrow \frac{d^n y}{dt^n} + a_{n-1}\frac{d^{n-1}y}{dt^{n-1}} + \dots + a_0 y = b_0 u$$

Then, convert this *n*th-order differential equation to *n* simultaneous first-order differential equations.



Find the state-space representation in phase-variable form.





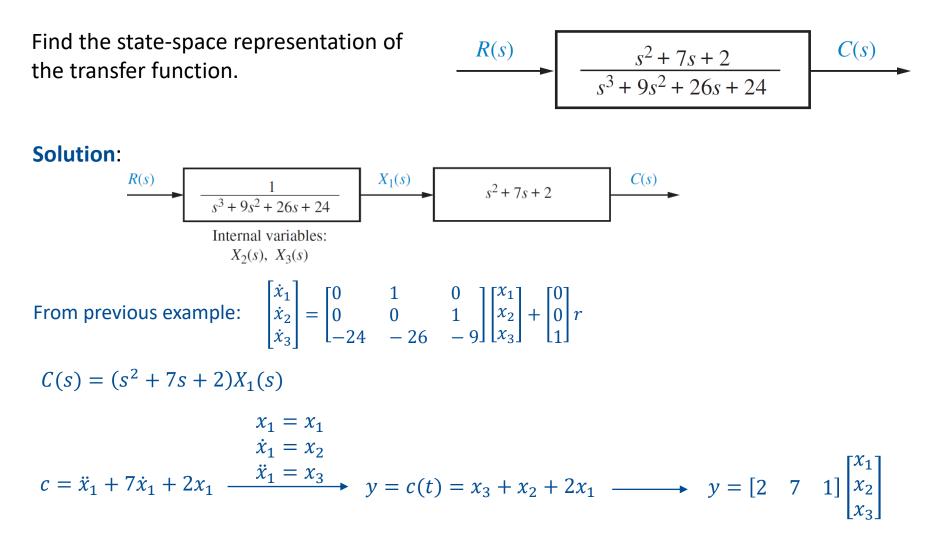
Converting a TF to SS

If a transfer function has a polynomial in *s* in the numerator, separate the transfer function into two cascaded transfer functions; the first is the denominator and the second is just the numerator.

$$\underbrace{\frac{U(s)}{s^{n} + a_{n-1}s^{n-1} + \dots + a_{0}}}_{W(s)} \underbrace{\frac{b_{m}s^{m} + b_{m-1}s^{n-1} + \dots + a_{0}}{s^{n} + a_{n-1}s^{n-1} + \dots + a_{0}}}_{Y(s)} \underbrace{\frac{U(s)}{s^{n} + a_{n-1}s^{n-1} + \dots + a_{0}}}_{w^{n} + b_{m-1}s^{m-1} + \dots + b_{0}} \underbrace{\frac{Y(s)}{s^{n} + a_{n-1}s^{n-1} + \dots + a_{0}}}_{w^{n} + b_{m-1}s^{m-1} + \dots + b_{0}} \underbrace{\frac{Y(s)}{s^{n} + a_{n-1}s^{n-1} + \dots + a_{0}}}_{w^{n} + b_{m-1}s^{m-1} + \dots + b_{0}} \underbrace{\frac{Y(s)}{s^{n} + a_{n-1}s^{n-1} + \dots + a_{0}}}_{w^{n} + b_{m-1}s^{m-1} + \dots + b_{0}} \underbrace{\frac{Y(s)}{s^{n} + a_{n-1}s^{n-1} + \dots + a_{0}}}_{w^{n} + b_{m-1}s^{m-1} + \dots + b_{0}} \underbrace{\frac{Y(s)}{s^{n} + a_{n-1}s^{n-1} + \dots + a_{0}}}_{w^{n} + b_{m-1}s^{m-1} + \dots + b_{0}} \underbrace{\frac{Y(s)}{s^{n} + a_{n-1}s^{n-1} + \dots + a_{0}}}_{w^{n} + b_{m-1}s^{m-1} + \dots + b_{0}} \underbrace{\frac{Y(s)}{s^{n} + a_{n-1}s^{n-1} + \dots + a_{0}}}_{w^{n} + b_{m-1}s^{m-1} + \dots + b_{0}} \underbrace{\frac{Y(s)}{s^{n} + a_{n-1}s^{n-1} + \dots + a_{0}}}_{w^{n} + b_{m-1}s^{m-1} + \dots + b_{0}}$$

- The first transfer function with just the denominator is converted to the phase-variable representation in state space.
- The second transfer function with just the numerator yields the output equation.





		State-Space Representation	MATLAB
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Using MATLAB and Control System Toolbox

Transfer-Function Representation Using tf, zpk

```
%% Transfer-Function Representation, Method 1
F1 = tf([3], [1 2 5 0]); \% \text{ or}
F1 = tf([3], conv([10], [125]));
F2 = tf([1], [10]);
F3 = tf([10], [1]);
F4 = tf(16, poly([0 - 1 - 1]));
%% Transfer-Function Representation, Method 2
s = tf('s');
F1 = 3/(s^3 + 2*s^2 + 5*s);
F2 = 1/s;
F3 = s:
F4 = \frac{16}{(s^*(s+1)^2)};
%% Transfer-Function Representation, Method 3
F1 = zpk([],[0-1+2i-1-2i],[3]);
```

```
F2 = zpk([],[0],[1]);
F3 = zpk([0],[],[1]);
```

```
F4 = zpk([],[0 -1 -1],[16]);
```

```
Method 1
```

Method 2

Method 3

sys = tf(numerator,denominator)

numerator and denominator are row vectors of polynomial coefficients in order of descending power.

```
F_1(s) = \frac{3}{s(s^2 + 2s + 5)}F_2(s) = \frac{1}{s}F_3(s) = sF_4(s) = \frac{16}{s(s+1)^2}
```

sys = zpk(zeros,poles,gain)

zero-pole-gain model with zeros and poles specified as row vectors of roots of numerator and denominator, and the scalar value of gain.

State-Space Representation Using ss and Conversions Using tf2ss, ss2tf

%% State-Space Representation A = [-4 - 1.5; 40];B = [2 0]'; $\dot{\mathbf{x}} = \begin{bmatrix} -4 & -1.5\\ 4 & 0 \end{bmatrix} \mathbf{x} + \begin{bmatrix} 2\\ 0 \end{bmatrix} u(t)$ $C = [1.5 \ 0.625];$ $v = [1.5 \quad 0.625]x$ D = 0;T ss = ss(A,B,C,D); % converting SS to TF $T(s) = \frac{3s+5}{s^2+4s+6}$ T tf = tf(T ss); %% TF to SS, SS to TF num = [172];den = [1 9 26 24]; [A, B, C, D] = tf2ss(num, den);T1 = ss(A,B,C,D); $T_1(s) = T_2(s) = \frac{s^2 + 7s + 2}{s^3 + 9s^2 + 26s + 24}$ T1 = tf(T1);% For SISO systems [num,den] = ss2tf(A,B,C,D); T2 = tf(num,den);