State-Space and System Modeling

State Space for Dynamical Systems

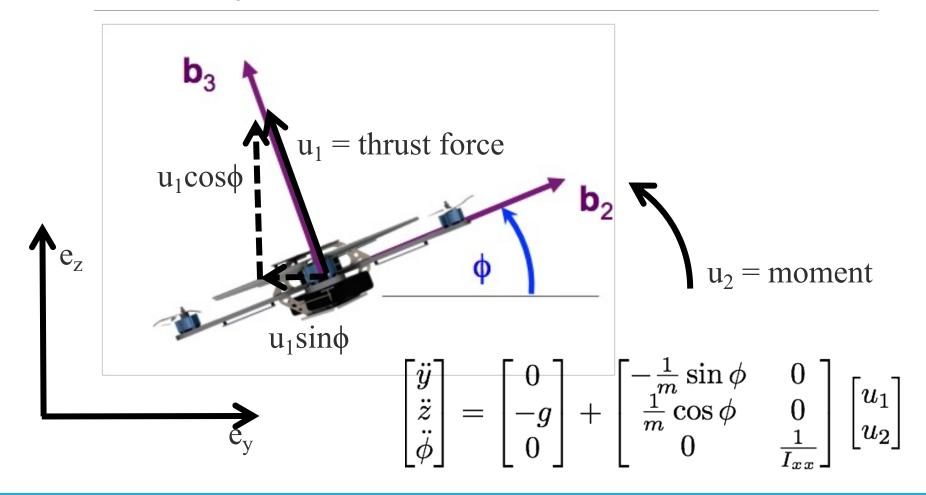
- \triangleright State: $\mathbf{x} = [x_1 \ x_2 \ \cdots \ x_n]^T$
- \triangleright Robotic systems: The state typically includes the configuration \mathbf{q} (position) and its derivative $\dot{\mathbf{q}}$ (velocity), that is

$$\mathbf{q} = \begin{bmatrix} q_1 \\ q_2 \\ \vdots \\ q_j \end{bmatrix}, \qquad \mathbf{x} = \begin{bmatrix} \mathbf{q} \\ \dot{\mathbf{q}} \end{bmatrix}$$

- The evolution of system's state over time is governed by a set of ordinary differential equations (ODEs).
- > ODEs are often expressed in their equivalent state-space form

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

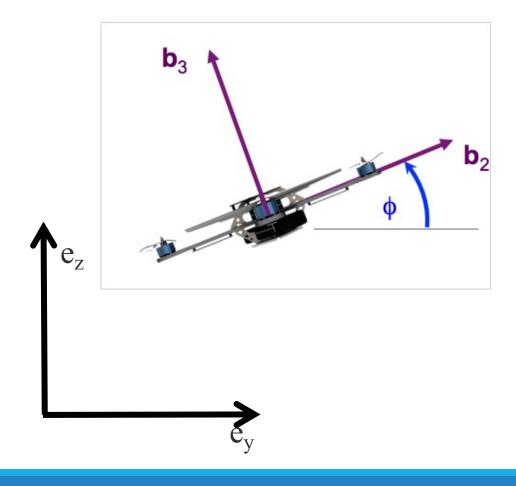
Example: Planar Quadrotor



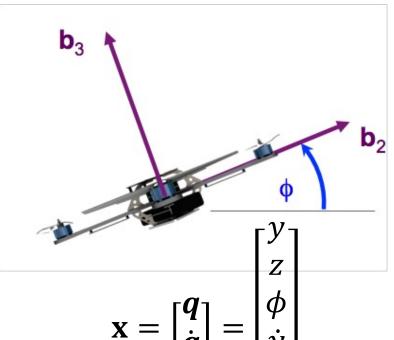
State Space

State vector

$$\mathbf{q} = \begin{bmatrix} y \\ z \\ \phi \end{bmatrix}, \mathbf{x} = \begin{bmatrix} \mathbf{q} \\ \dot{\mathbf{q}} \end{bmatrix}$$



Planar Quadrotor Model



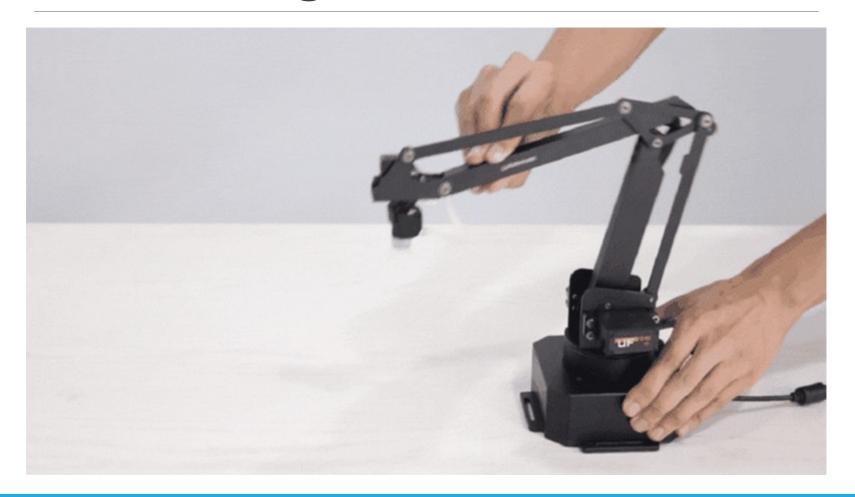
$$\mathbf{x} = \begin{bmatrix} \boldsymbol{q} \\ \dot{\boldsymbol{q}} \end{bmatrix} = \begin{bmatrix} z \\ \phi \\ \dot{y} \\ \dot{z} \\ \dot{\phi} \end{bmatrix}$$

$$\begin{bmatrix} \ddot{y} \\ \ddot{z} \\ \ddot{\phi} \end{bmatrix} = \begin{bmatrix} 0 \\ -g \\ 0 \end{bmatrix} + \begin{bmatrix} -\frac{1}{m}\sin\phi & 0 \\ \frac{1}{m}\cos\phi & 0 \\ 0 & \frac{1}{I_{xx}} \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix}$$

$$\mathbf{x} = \begin{bmatrix} \mathbf{q} \\ \dot{\mathbf{q}} \end{bmatrix} = \begin{bmatrix} \mathbf{y} \\ \dot{\mathbf{y}} \\ \dot{\dot{\mathbf{z}}} \\ \dot{\phi} \end{bmatrix} \qquad \dot{\mathbf{x}} = \begin{bmatrix} x_4 \\ x_5 \\ x_6 \\ 0 \\ -g \\ 0 \end{bmatrix} + \begin{bmatrix} 0 & 0 \\ 0 & 0 \\ -m^{-1} \sin x_3 & 0 \\ m^{-1} \cos x_3 & 0 \\ 0 & I_{xx}^{-1} \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix}$$

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

A Modeling Choice



A Modeling Choice



Main Steps to Build a State-Space Model

Given an ODE (for now of a single variable, y(t))

- \triangleright Isolate the n^{th} highest derivative, $y^{(n)} = g(y, \dot{y}, ..., y^{(n-1)}, \mathbf{u})$
- \triangleright Set $x_1 = y(t)$, $x_2 = \dot{y}(t)$, ..., $x_n = y^{(n-1)}(t)$
- \triangleright Create state vector $\mathbf{x} = \begin{bmatrix} x_1 & x_2 & \cdots & x_n \end{bmatrix}^T = \begin{bmatrix} y & \dot{y} & \cdots & y^{(n-1)} \end{bmatrix}^T$
- > Rewrite into a system of coupled first-order differential equations

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

$$\dot{x}_2 = \ddot{y} = x_3$$

$$\dot{x}_n = y^{(n)} = g(y, \dot{y}, ..., y^{(n-1)}, \mathbf{u}) = g(x_1, x_2, ..., x_n, \mathbf{u})$$

Main Steps to Build a State-Space Model

> Rewrite in matrix form

$$\begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \\ \vdots \\ \dot{x}_n \end{bmatrix} = \begin{bmatrix} x_2 \\ x_3 \\ \vdots \\ g(x_1, x_2, \dots, x_n, \mathbf{u}) \end{bmatrix}$$

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

$$\mathbf{x} \in \mathbb{R}^n$$
, $\mathbf{u} \in \mathbb{R}^m$

n states m inputs

➤ Note: A system is linear time-invariant (LTI) when

$$\dot{\mathbf{x}} = A\mathbf{x} + B\mathbf{u},$$

and A is an $n \times n$ constant matrix, and B is an $n \times m$ constant matrix.

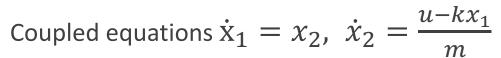
Example: Spring-Mass System

$$m\ddot{q}(t) + kq(t) = u(t)$$

Order of the system = 2

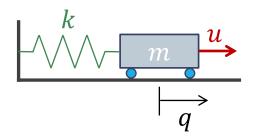
Rewrite
$$\ddot{q}(t) = \frac{u(t) - kq(t)}{m}$$

State vector
$$\mathbf{x} = [x_1 \ x_2]^{\mathrm{T}} = [q \ \dot{q}]^{\mathrm{T}}$$



$$\operatorname{Then}\begin{bmatrix}\dot{x}_1\\\dot{x}_2\end{bmatrix} = \begin{bmatrix} x_2\\\underline{u-kx_1}\\m \end{bmatrix}, \ \operatorname{or}\ \begin{bmatrix}\dot{x}_1\\\dot{x}_2\end{bmatrix} = \begin{bmatrix} 0&1\\-\frac{k}{m}&0 \end{bmatrix}\begin{bmatrix}x_1\\x_2\end{bmatrix} + \begin{bmatrix}0\\\frac{1}{m}\end{bmatrix}u$$

> Linear system



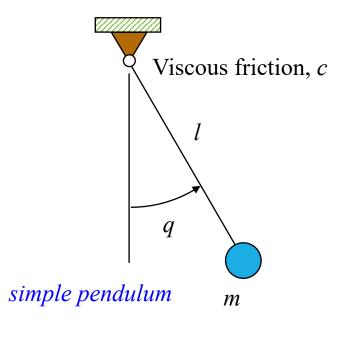
Example: Damped Pendulum

Equation of motion

$$\ddot{q} + \frac{c}{ml^2}\dot{q} + \frac{g}{l}\sin q = 0$$

State space representation

$$x = \begin{bmatrix} q \\ \dot{q} \end{bmatrix}, \dot{x} = \begin{bmatrix} x_2 \\ -\frac{g}{l}\sin x_1 - \frac{c}{ml^2}x_2 \end{bmatrix} \quad \text{simple pendulum}$$



Nonlinear \implies Linearize around equilibria $\dot{\mathbf{x}} = f(\mathbf{x}) \equiv 0$

Example: Damped Pendulum

Equation of motion

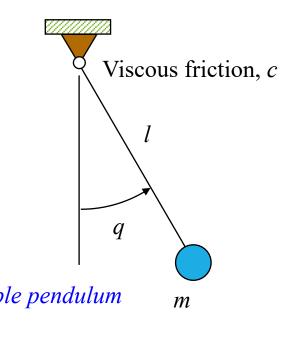
$$\ddot{q} + \frac{c}{ml^2}\dot{q} + \frac{g}{l}\sin q = 0$$

State space representation

$$x = \begin{bmatrix} q \\ \dot{q} \end{bmatrix}, \dot{x} = \begin{bmatrix} x_2 \\ -\frac{g}{l} \sin x_1 - \frac{c}{ml^2} x_2 \end{bmatrix} \quad \text{simple pendulum}$$

Equilibrium point(s)

$$x_{e,1} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}, x_{e,2} = \begin{bmatrix} \pi \\ 0 \end{bmatrix}$$



Consider points near equilibria

$$\widetilde{x} = (x - x_{e,i})$$

Equilibria

Consider a system with n degrees of freedom

Let q_e be a configuration at static equilibrium ($\dot{\mathbf{x}} = f(\mathbf{x}) \equiv 0$)

$$x(t_0) = \begin{bmatrix} q_e \\ \vdots \\ 0 \end{bmatrix} \Rightarrow x(t > t_0) = \begin{bmatrix} q_e \\ \vdots \\ 0 \end{bmatrix}$$

- > An equilibrium point can be
 - Stable
 - Unstable
 - Critically stable (or neutrally stable)
- ➤ We are interested in the behavior of the system around equilibrium points.
- ➤ We linearize the system around equilibria!

Linearization

- Fiven a nonlinear system $\dot{\mathbf{x}} = f(\mathbf{x}, \mathbf{u}), \mathbf{x}, f \in \mathbb{R}^n, \mathbf{u} \in \mathbb{R}^m$, derive an approximate linear system $\dot{\mathbf{x}} = A\mathbf{x} + B\mathbf{u}$ about an equilibrium point $(\mathbf{x}_e, \mathbf{u}_e)$.
- > Taylor series expansion around equilibrium point:

$$f(\mathbf{x}_{e} + \Delta \mathbf{x}, \mathbf{u}_{e} + \Delta \mathbf{u}) = f(\mathbf{x}_{e}, \mathbf{u}_{e}) + \left[\frac{\partial f}{\partial \mathbf{x}}\right]_{(\mathbf{x}_{e}, \mathbf{u}_{e})} \Delta \mathbf{x} + \left[\frac{\partial f}{\partial \mathbf{u}}\right]_{(\mathbf{x}_{e}, \mathbf{u}_{e})} \Delta \mathbf{u} + \mathbf{H.O.T.}$$

$$\dot{\mathbf{x}} + \Delta \dot{\mathbf{x}} \approx f(\mathbf{x}_{e}, \mathbf{u}_{e}) + \left[\frac{\partial f}{\partial \mathbf{x}}\right]_{(\mathbf{x}_{e}, \mathbf{u}_{e})} \Delta \mathbf{x} + \left[\frac{\partial f}{\partial \mathbf{u}}\right]_{(\mathbf{x}_{e}, \mathbf{u}_{e})} \Delta \mathbf{u}$$

$$\Rightarrow \Delta \dot{\mathbf{x}} = A\Delta \mathbf{x} + B\Delta \mathbf{u}$$

Re-defining $\Delta \mathbf{x} \triangleq \mathbf{x}$, and $\Delta \mathbf{u} \triangleq \mathbf{u}$ yields $\dot{\mathbf{x}} = A\mathbf{x} + B\mathbf{u}$ with

$$A_{n\times n} = \begin{bmatrix} \frac{\partial f}{\partial \mathbf{x}} \end{bmatrix}_{(\mathbf{x}_e, \mathbf{u}_e)} = \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \cdots & \frac{\partial f_1}{\partial x_n} \\ \vdots & \ddots & \vdots \\ \frac{\partial f_n}{\partial x_1} & \cdots & \frac{\partial f_n}{\partial x_n} \end{bmatrix}_{(\mathbf{x}_e, \mathbf{u}_e)}, B_{n\times m} = \begin{bmatrix} \frac{\partial f}{\partial \mathbf{u}} \end{bmatrix}_{(\mathbf{x}_e, \mathbf{u}_e)} = \begin{bmatrix} \frac{\partial f_1}{\partial u_1} & \cdots & \frac{\partial f_1}{\partial u_n} \\ \vdots & \ddots & \vdots \\ \frac{\partial f_n}{\partial u_1} & \cdots & \frac{\partial f_n}{\partial u_n} \end{bmatrix}_{(\mathbf{x}_e, \mathbf{u}_e)}$$

Pendulum Linearization

$$\dot{\mathbf{x}} = \left[-\frac{g}{l} \sin x_1 - \frac{c}{ml^2} x_2 \right] = f(\mathbf{x})$$

Equilibrium point 1

$$\begin{split} \mathbf{x}_{e,1} &= \begin{bmatrix} 0 \\ 0 \end{bmatrix} \\ \dot{\mathbf{x}} &\approx \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} \\ \frac{\partial f_2}{\partial x_1} & \frac{\partial f_2}{\partial x_2} \end{bmatrix}_{\mathbf{x}_{e,1}} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \\ &= \begin{bmatrix} 0 & 1 \\ -\frac{g}{l} & -\frac{c}{ml^2} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \end{split}$$

Equilibrium point 2

$$\begin{aligned} \mathbf{x}_{e,2} &= \begin{bmatrix} \pi \\ 0 \end{bmatrix} \\ \dot{\mathbf{x}} &\approx \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} \\ \frac{\partial f_2}{\partial x_1} & \frac{\partial f_2}{\partial x_2} \end{bmatrix}_{\mathbf{x}_{e,2}} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \\ &= \begin{bmatrix} 0 & 1 \\ \frac{g}{l} & -\frac{c}{ml^2} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \end{aligned}$$

Pendulum Stability

A linear system $\dot{x} = Ax$ is stable iff the real parts of all eigenvalues of A are negative.

Equilibrium point 1

$$\mathbf{x}_{e,1} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$\dot{\mathbf{x}} \approx \begin{bmatrix} 0 & 1 \\ -\frac{g}{l} & -\frac{c}{ml^2} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

$$\lambda = -\frac{c}{2ml^2} \pm \sqrt{\left(\frac{c}{2ml^2}\right)^2 - \frac{g}{l}}$$

Stable

Marginally stable if c = 0

Equilibrium point 2

$$\mathbf{x}_{e,2} = \begin{bmatrix} \pi \\ 0 \end{bmatrix}$$

$$\dot{\mathbf{x}} \approx \begin{bmatrix} 0 & 1 \\ \frac{g}{l} & -\frac{c}{ml^2} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

$$\lambda = -\frac{c}{2ml^2} \pm \sqrt{\left(\frac{c}{2ml^2}\right)^2 + \frac{g}{l}}$$

Unstable