

# Lecture 4: Stability Analysis

## Underactuated Robotics

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# Agenda



## Introduction

## Inverted Pendulum

- Modelling

- Linearization

- Stability Analysis

- Controllability Analysis

- Stabilization using State Feedback

## Acrobot

- Modelling

- Linearization

- Controllability Analysis

- Stabilization using State Feedback

## Stability Analysis

- Global Stability Analysis

- Local Stability Analysis



### Knowledge:

- ▶ Derive dynamical state-space models of robots as control systems
- ▶ **Analyze the stability of low dimensional linear and nonlinear systems**
- ▶ **Analyze the** observability and **controllability of linear control systems**
- ▶ Use a variety of controllers for underactuated robots

### Skills:

- ▶ Implement simulations of control systems in software
- ▶ Create concise technical reports presenting solutions to proposed problems

### Competencies:

- ▶ Choose appropriate modern control techniques to solve control problems in robotics
- ▶ Apply modern control techniques to control simulated underactuated robots



- ▶ **Lesson 1:** Newton-Euler Modelling
- ▶ **Lesson 2:** Euler-Lagrange Modelling
- ▶ **Lesson 3:** Simulation of Robot Dynamics
- ▶ **Lesson 4:** Stability Analysis
- ▶ **Lesson 5:** Optimal Control
- ▶ **Lesson 6:** Energy Shaping Control
- ▶ **Lesson 7:** Feedback Linearisation
- ▶ **Lesson 8:** Sliding Mode Control
- ▶ **Lesson 9:** Simulation and Implementation of Control Systems
- ▶ **Lesson 10:** Optimization-Based Control
- ▶ **Lesson 11:**
- ▶ **Lesson 12:**

# Inverted Pendulum



## Introduction

## Inverted Pendulum

- Modelling

- Linearization

- Stability Analysis

- Controllability Analysis

- Stabilization using State Feedback

## Acrobot

- Modelling

- Linearization

- Controllability Analysis

- Stabilization using State Feedback

## Stability Analysis

- Global Stability Analysis

- Local Stability Analysis



## Introduction

## Inverted Pendulum

Modelling

Linearization

Stability Analysis

Controllability Analysis

Stabilization using State Feedback

## Acrobot

Modelling

Linearization

Controllability Analysis

Stabilization using State Feedback

## Stability Analysis

Global Stability Analysis

Local Stability Analysis



If  $q$  is a trajectory of a mechanical system that is affected by a generalized force  $Q$  then

$$\frac{d}{dt} \frac{\partial \mathcal{L}}{\partial \dot{q}} - \frac{\partial \mathcal{L}}{\partial q} = Q$$

where  $Q$  is an  $n$ -dimensional vector of generalized forces.



The ***potential energy*** of the inverted pendulum is

$$E_{\text{pot}} = mgl(1 + \cos(\theta)) \quad [\text{J}]$$

where  $m$  is the mass of the pendulum [kg],  $l$  is the length from the pivot point to the center of mass [m], and  $g$  is the gravitational acceleration [ $\text{m/s}^2$ ].





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The ***kinetic energy*** is

$$E_{\text{kin}} = \frac{1}{2}mv^T v + \frac{1}{2}\omega^T I \omega$$

where  $v$  is the velocity of the center of mass [ $\text{m/s}$ ],  $\omega$  is the angular velocity of the center of mass [ $\text{rad/s}$ ], and  $I$  is the moment of inertia of the pendulum with respect to the center of mass (given in Base frame) [ $\text{kgm}^2$ ].



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$$E_{\text{kin}} = \frac{1}{2}ml^2\omega^2$$



By application of Lagrange–D'Alembert's Principle

$$ml^2\dot{\omega} - mgl \sin \theta = -b\omega + \tau$$

which leads to the following equation of motion

$$\dot{\omega} = \frac{g}{l} \sin \theta - \frac{b}{ml^2} \omega + \frac{1}{ml^2} \tau$$

# Modelling

## Equation of Motion on State Space Form



We rewrite the equation of motion (with generalized coordinate  $q = \theta$ )

$$\dot{\omega} = \frac{g}{l} \sin \theta - \frac{b}{ml^2} \omega + \frac{1}{ml^2} \tau$$

on state space form

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

by defining the state and input

$$x = \begin{bmatrix} \theta \\ \omega \end{bmatrix}, \quad u = \tau$$

# Modelling

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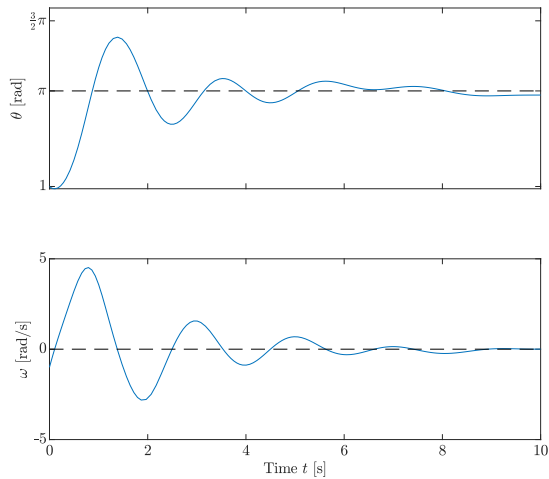
$$x = \begin{bmatrix} \theta \\ \omega \end{bmatrix}, \quad u = \tau$$

This implies that

$$\dot{x} = \begin{bmatrix} \dot{\theta} \\ \dot{\omega} \end{bmatrix} = \underbrace{\begin{bmatrix} \omega \\ \frac{g}{l} \sin \theta - \frac{b}{ml^2} \omega + \frac{1}{ml^2} \tau \end{bmatrix}}_{=f(x,u)}$$

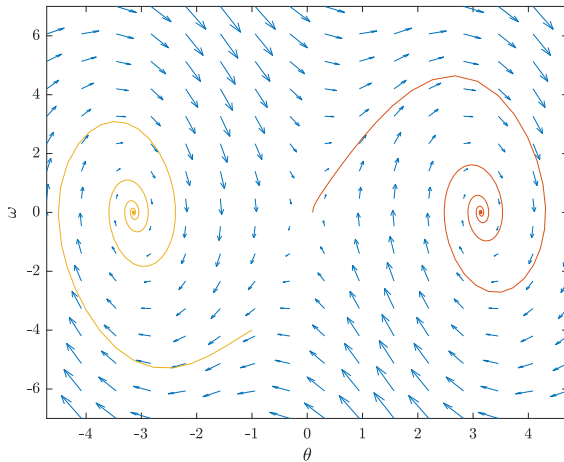


```
1 %% Simulation of Pendulum
2 t0 = 0;           % [s] Start time for simulation
3 tEnd = 50;        % [s] End time for simulation
4 tu = (t0:0.05:tEnd)';
5 u = cos(tu);      % Specification of input
6 x0 = [1;-1];      % Specification of initial state
7 % Simulate the system
8 [t,x] = ode45(@invertedPendulum,[t0 tEnd],x0,[],[tu u]);
9 %% Function used for simulation via ode45
10 function dx = invertedPendulum(t,x,u)
11     u = interp1(u(:,1),u(:,2),t); % determine u by linear interpolation
12     theta = x(1); omega = x(2); tau = u;
13     dx = [omega;
14           g/l*sin(theta) - b/(l^2*m)*omega+1/(l^2*m)*tau];
15 end
```



# Modelling

## Phase Portrait





# Inverted Pendulum

Linearization



## Introduction

## Inverted Pendulum

Modelling

Linearization

Stability Analysis

Controllability Analysis

Stabilization using State Feedback

## Acrobot

Modelling

Linearization

Controllability Analysis

Stabilization using State Feedback

## Stability Analysis

Global Stability Analysis

Local Stability Analysis

# Linearization

## First Order Taylor Approximation



The system of 1st order differential equations

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

is approximated at  $x = \bar{x}$  using first-order Taylor approximation by

$$\dot{x} \approx f(\bar{x}, \bar{u}) + \left. \frac{\partial f}{\partial x} \right|_{x=\bar{x}, u=\bar{u}} \hat{x} + \left. \frac{\partial f}{\partial u} \right|_{x=\bar{x}, u=\bar{u}} \hat{u}$$

where  $\hat{x} = x - \bar{x}$  and all partial derivatives are evaluated at the operating point  $p = (\bar{x}, \bar{u})$ .

# Linearization

## First Order Taylor Approximation



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where  $\hat{x} = x - \bar{x}$  and all partial derivatives are evaluated at the operating point  $p = (\bar{x}, \bar{u})$ .

We linearize systems at equilibrium points, i.e., where  $\dot{x} = 0$ . This implies that

$$\dot{\hat{x}} \approx \underbrace{\left. \frac{\partial f}{\partial x} \right|_{x=\bar{x}, u=\bar{u}}}_{=A} \hat{x} + \underbrace{\left. \frac{\partial f}{\partial u} \right|_{x=\bar{x}, u=\bar{u}}}_{B} \hat{u}$$



The dynamics of the inverted pendulum is given by

$$\begin{bmatrix} \dot{\theta} \\ \dot{\omega} \end{bmatrix} = \begin{bmatrix} \omega \\ \frac{g}{l} \sin \theta - \frac{b}{ml^2} \omega + \frac{1}{ml^2} \tau \end{bmatrix}$$

This implies that

$$\frac{\partial f}{\partial(\theta, \omega)} = A = \begin{bmatrix} \frac{\partial f_1}{\partial \theta} & \frac{\partial f_1}{\partial \omega} \\ \frac{\partial f_2}{\partial \theta} & \frac{\partial f_2}{\partial \omega} \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ \frac{g}{l} \cos \theta & -\frac{b}{ml^2} \end{bmatrix}$$

and

$$\frac{\partial f}{\partial \tau} = B = \begin{bmatrix} \frac{\partial f_1}{\partial \tau} \\ \frac{\partial f_2}{\partial \tau} \end{bmatrix} = \begin{bmatrix} 0 \\ \frac{1}{ml^2} \end{bmatrix}$$



The dynamics of the inverted pendulum is given by

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and

$$\frac{\partial f}{\partial \tau} = B = \begin{bmatrix} \frac{\partial f_1}{\partial \tau} \\ \frac{\partial f_2}{\partial \tau} \end{bmatrix} = \begin{bmatrix} 0 \\ \frac{1}{ml^2} \end{bmatrix}$$

At an equilibrium point  $\dot{x} = f(\bar{x}, \bar{u}) = 0$ , which means that

$$\bar{\omega} = 0, \quad \bar{\tau} = mlg \sin(\bar{\theta})$$



The linearized model of the inverted pendulum at equilibrium point  $(\bar{x}, \bar{u}) = (\bar{\theta}, \bar{\omega}, \bar{\tau}) = (\bar{\theta}, 0, mlg \sin(\bar{\theta}))$  is given by

$$\dot{\hat{x}} = \begin{bmatrix} \dot{\hat{\theta}} \\ \dot{\hat{\omega}} \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ \frac{g}{l} \cos \bar{\theta} & -\frac{1}{ml^2} \end{bmatrix} \underbrace{\begin{bmatrix} \hat{\theta} \\ \hat{\omega} \end{bmatrix}}_{=\hat{x}} + \begin{bmatrix} 0 \\ \frac{1}{ml^2} \end{bmatrix} \hat{u}$$

where  $\hat{x} = (x - \bar{x})$  and  $\hat{u} = (u - \bar{u})$ .

# Linearization

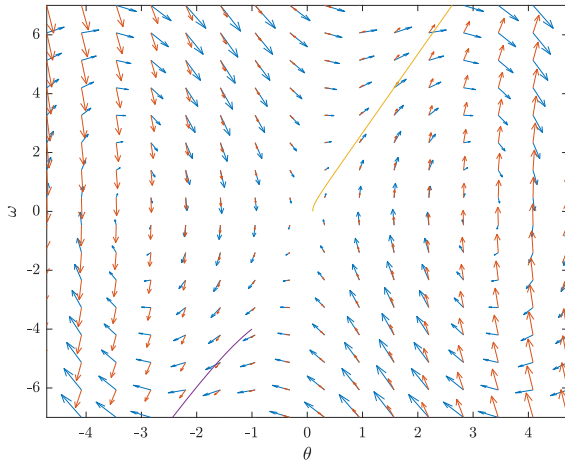
## MATLAB Code for Simulation



```
1 %% Simulation of Linearized Pendulum
2 pendul_lin = ss(A,B,C,D); % Definition of system model
3 t0 = 0;           % [s] Start time for simulation
4 tEnd = 50;        % [s] End time for simulation
5 t = (t0:0.05:tEnd)';
6 u = cos(t);       % Specification of input
7 x0 = [1;-1];      % Specification of initial state
8 % Simulate the system
9 [y,t,x] = lsim(pendul_lin,u,t,x0);
```

# Linearization

Simulation of Linear System





# Inverted Pendulum

Stability Analysis



## Introduction

## Inverted Pendulum

Modelling

Linearization

Stability Analysis

Controllability Analysis

Stabilization using State Feedback

## Acrobot

Modelling

Linearization

Controllability Analysis

Stabilization using State Feedback

## Stability Analysis

Global Stability Analysis

Local Stability Analysis



A linear continuous-time system described by the state equation

$$\dot{x} = Ax$$

is asymptotically stable if and only if all eigenvalues have negative real part.

# Stability Analysis

## Stability of Linearized Pendulum



The eigenvalues of the system matrix (matrix  $A$ ) of the linearized pendulum system are the values of  $\lambda$  that solves the equation (eigenvalues of  $A$ )

$$\det(\lambda I - A) = 0$$



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The eigenvalues are given by

$$\det \left( \begin{bmatrix} \lambda & -1 \\ -\frac{g}{l} \cos \bar{\theta} & \lambda + \frac{b}{ml^2} \end{bmatrix} \right) = \lambda \left( \lambda + \frac{b}{ml^2} \right) - \frac{g}{l} \cos \bar{\theta} = 0$$

The eigenvalues are therefore

$$\lambda = \frac{-\frac{b}{ml^2} \pm \sqrt{\left(\frac{b}{ml^2}\right)^2 + 4\frac{g}{l} \cos \bar{\theta}}}{2}$$



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In conclusion, one pole will be in the left half-plane, when  $\cos(\bar{\theta}) > 0$ ; when  $\cos(\bar{\theta}) < 0$  then the linearized system is stable.

# Stability Analysis

Relation between Stability of Linearized Pendulum and Nonlinear Pendulum



By Hartman-Grobman Theorem, if the poles of a linearized system are not on the imaginary axis, then the nonlinear and linearized models behave similarly in a small neighborhood about the equilibrium point.

# Stability Analysis

Relation between Stability of Linearized Pendulum and Nonlinear Pendulum



By Hartman-Grobman Theorem, if the poles of a linearized system are not on the imaginary axis, then the nonlinear and linearized models behave similarly in a small neighborhood about the equilibrium point.

**For us:** Design a stabilizing controller for the linearized system model, then the controller also stabilizes the nonlinear system around the equilibrium point.

# Inverted Pendulum

## Controllability Analysis



## Introduction

## Inverted Pendulum

Modelling

Linearization

Stability Analysis

**Controllability Analysis**

Stabilization using State Feedback

## Acrobot

Modelling

Linearization

Controllability Analysis

Stabilization using State Feedback

## Stability Analysis

Global Stability Analysis

Local Stability Analysis





A continuous time system

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

is said to be *controllable* iff for any  $\xi \in \mathbb{R}^n$  there exists  $u(t)$  such that for some  $T > 0$ ,  $x(T) = \xi$ .

A system

$$\dot{x}(t) = Ax(t) + Bu(t)$$

where  $x \in \mathbb{R}^n$ ,  $u \in \mathbb{R}^m$ , is controllable if and only if

$$\text{rank} \begin{bmatrix} B & AB & \dots & A^{n-1}B \end{bmatrix} = n$$



For a state space model

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

a *state feedback* is a feedback of the form

$$u = Fx$$

Combining these two equations, we obtain:

$$\dot{x} = Ax + BFx = (A + BF)x$$

Thus, the result of a state feedback is a system with a modified system matrix, and thus with modified poles.



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Thus, the result of a state feedback is a system with a modified system matrix, and thus with modified poles.

The poles of  $A_{cl} = A + BF$  can be placed anywhere in the complex plane if the system is controllable.



The controllability matrix of the linearized pendulum is

$$\mathcal{C} = [B \quad AB] = \begin{bmatrix} 0 & \frac{1}{ml^2} \\ \frac{1}{ml^2} & -\frac{b}{m^2 l^4} \end{bmatrix}$$

It is seen that the system is controllable, as  $\mathcal{C}$  has full rank.

# Inverted Pendulum

Stabilization using State Feedback



## Introduction

## Inverted Pendulum

Modelling

Linearization

Stability Analysis

Controllability Analysis

Stabilization using State Feedback

## Acrobot

Modelling

Linearization

Controllability Analysis

Stabilization using State Feedback

## Stability Analysis

Global Stability Analysis

Local Stability Analysis

# Stabilization using State Feedback

State Feedback of Linearized System



To stabilize the pendulum in upright, a feedback controller needs to be designed; it is possible since the linearized system is controllable. Poles must be placed in the open left-half plane, and the system will respond faster when the poles are more negative.

# Stabilization using State Feedback

## State Feedback of Linearized System



To stabilize the pendulum in upright, a feedback controller needs to be designed; it is possible since the linearized system is controllable. Poles must be placed in the open left-half plane, and the system will respond faster when the poles are more negative.

A controller for the pendulum with poles  $s = -1$  and  $s = -2$  is

$$u = Fx$$

where

$$F = [-11.82 \quad -2]$$

# Stabilization using State Feedback

MATLAB Code for Pole Placement

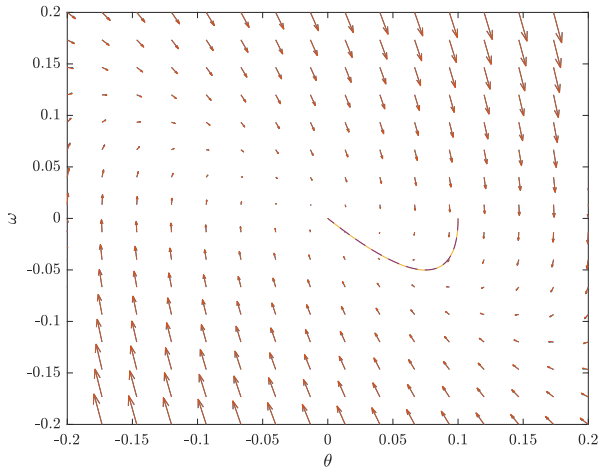


```
1 %% Pole Placement for Linearized Pendulum
2 pendul_lin = ss(A,B,C,D); % Definition of system model
3 P = [-1 -2];
4 F = -place(pendul_lin.A, pendul_lin.B,P);
```



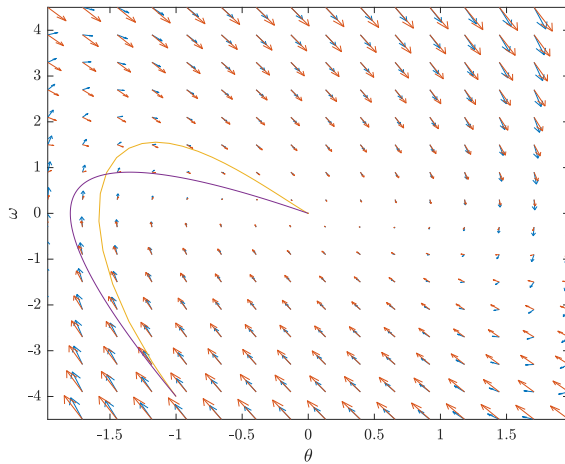
# Stabilization using State Feedback

Phase Plot (Initial Condition Close to (0,0))



# Stabilization using State Feedback

Phase Plot (Initial Condition Far from (0,0))





Introduction

Inverted Pendulum

Modelling

Linearization

Stability Analysis

Controllability Analysis

Stabilization using State Feedback

**Acrobot**

Modelling

Linearization

Controllability Analysis

Stabilization using State Feedback

Stability Analysis

Global Stability Analysis

Local Stability Analysis

## Introduction

## Inverted Pendulum

Modelling

Linearization

Stability Analysis

Controllability Analysis

Stabilization using State Feedback

## Acrobot

Modelling

Linearization

Controllability Analysis

Stabilization using State Feedback

## Stability Analysis

Global Stability Analysis

Local Stability Analysis



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where  $Q$  is an  $n$ -dimensional vector of generalized forces.



The ***potential energy*** of the inverted pendulum is

$$E_{\text{pot}} = g(m_1(-l_{c1} \cos \theta_1) + m_2(-l_1 \cos \theta_1 - l_{c2} \cos(\theta_1 + \theta_2))) \quad [\text{J}]$$

where  $m$  is the mass of the pendulum [kg],  $l_{ci}$  is the length from the  $i$ th joint to the  $i$ th center of mass [m], and  $g$  is the gravitational acceleration [m/s<sup>2</sup>].

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The **kinetic energy** is

$$E_{\text{kin}} = \frac{1}{2} \begin{bmatrix} \dot{\theta}_1 & \dot{\theta}_2 \end{bmatrix} \begin{bmatrix} I_1 + I_2 + m_2 l_1^2 + 2m_2 l_1 l_{c2} \cos \theta_2 & I_2 + m_2 l_1 l_{c2} \cos \theta_2 \\ I_2 + m_2 l_1 l_{c2} \cos \theta_2 & I_2 \end{bmatrix} \begin{bmatrix} \dot{\theta}_1 \\ \dot{\theta}_2 \end{bmatrix}$$

where  $v$  is the velocity of the center of mass [ $\text{m/s}$ ],  $\omega$  is the angular velocity of the center of mass [ $\text{rad/s}$ ], and  $I$  is the moment of inertia of the pendulum with respect to the center of mass (given in Base frame) [ $\text{kgm}^2$ ].

By application of Lagrange–D'Alembert's Principle ( $q = (\theta_1, \theta_2)$ )

$$M(q)\ddot{q} + C(q, \dot{q})\dot{q} + g(q) = B\tau$$

where

$$M(q) = \begin{bmatrix} I_1 + I_2 + m_2 l_1^2 + 2m_2 l_1 l_{c2} \cos \theta_2 & I_2 + m_2 l_1 l_{c2} \cos \theta_2 \\ I_2 + m_2 l_1 l_{c2} \cos \theta_2 & I_2 \end{bmatrix}$$
$$C(q, \dot{q}) = \begin{bmatrix} -2m_2 l_1 l_{c2} \sin \theta_2 \dot{q}_2 & -m_2 l_1 l_{c2} \sin \theta_2 \dot{q}_2 \\ m_2 l_1 l_{c2} \sin \theta_2 \dot{q}_1 & 0 \end{bmatrix}$$
$$g(q) = \begin{bmatrix} g(m_1(l_{c1} \sin \theta_1) + m_2(l_1 \sin \theta_1 + l_{c2} \sin(\theta_1 + \theta_2))) \\ gm_2 l_{c2} \sin(\theta_1 + \theta_2) \end{bmatrix}$$
$$B = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$





We rewrite the equation of motion (with generalized coordinate  $q = \theta$ )

$$\ddot{q} = \begin{bmatrix} \dot{\omega}_1 \\ \dot{\omega}_2 \end{bmatrix} = M^{-1}(q) (B\tau - C(q, \dot{q})\dot{q} - g(q))$$

on state space form

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

by defining the state and input

$$x = [\theta_1 \quad \theta_2 \quad \omega_1 \quad \omega_2]^T, \quad u = \tau$$



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on state space form

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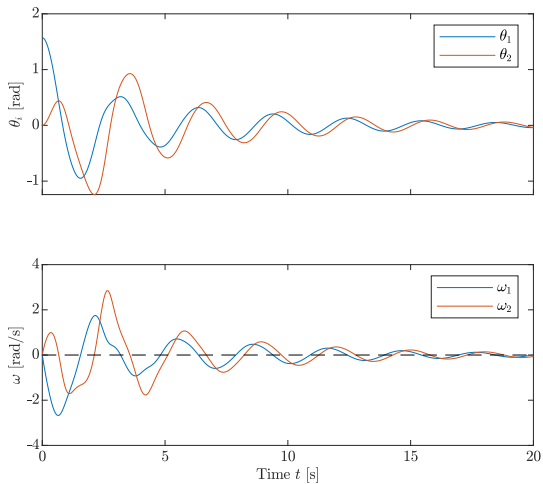
$$x = [\theta_1 \quad \theta_2 \quad \omega_1 \quad \omega_2]^T, \quad u = \tau$$

This implies that

$$\dot{x} = \begin{bmatrix} \dot{\theta}_1 \\ \dot{\theta}_2 \\ \dot{\omega}_1 \\ \dot{\omega}_2 \end{bmatrix} = \begin{bmatrix} \omega_1 \\ \omega_2 \\ \underbrace{M^{-1}(q) (B\tau - C(q, \dot{q})\dot{q} - g(q))}_{=f(x,u)} \end{bmatrix}$$

# Modelling

## Simulation of Acrobot



## Introduction

## Inverted Pendulum

Modelling

Linearization

Stability Analysis

Controllability Analysis

Stabilization using State Feedback

## Acrobot

Modelling

Linearization

Controllability Analysis

Stabilization using State Feedback

## Stability Analysis

Global Stability Analysis

Local Stability Analysis

# Linearization

## First Order Taylor Approximation



The system of 1st order differential equations

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$$\dot{x} \approx f(\bar{x}, \bar{u}) + \left. \frac{\partial f}{\partial x} \right|_{x=\bar{x}, u=\bar{u}} \hat{x} + \left. \frac{\partial f}{\partial u} \right|_{x=\bar{x}, u=\bar{u}} \hat{u}$$

where  $\hat{x} = x - \bar{x}$  and all partial derivatives are evaluated at the operating point  $p = (\bar{x}, \bar{u})$ .

# Linearization

## First Order Taylor Approximation



The system of 1st order differential equations

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

is approximated at  $x = \bar{x}$  using first-order Taylor approximation by

$$\dot{x} \approx f(\bar{x}, \bar{u}) + \left. \frac{\partial f}{\partial x} \right|_{x=\bar{x}, u=\bar{u}} \hat{x} + \left. \frac{\partial f}{\partial u} \right|_{x=\bar{x}, u=\bar{u}} \hat{u}$$

where  $\hat{x} = x - \bar{x}$  and all partial derivatives are evaluated at the operating point  $p = (\bar{x}, \bar{u})$ .

We linearize systems at equilibrium points, i.e., where  $\dot{x} = 0$ . This implies that

$$\dot{\hat{x}} \approx \underbrace{\left. \frac{\partial f}{\partial x} \right|_{x=\bar{x}, u=\bar{u}}}_{=A} \hat{x} + \underbrace{\left. \frac{\partial f}{\partial u} \right|_{x=\bar{x}, u=\bar{u}}}_{B} \hat{u}$$

# Linearization

## Operating Point



The Acrobot is linearized at an equilibrium point  $\dot{x} = f(\bar{x}, \bar{u}) = 0$ ; in particular, at its upright position which means that

$$\bar{\theta}_1 = \pi, \bar{\theta}_2 = 0, \bar{\omega}_1 = \bar{\omega}_2 = 0, \quad \bar{\tau} = 0$$



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The linearized system has poles:  $s = -4.6882, s = -2.2014, s = 2.2628, s = 1.7697$ .



# Linearization

MATLAB Code for Linearization



```
1 %% Linearization of Acrobot
2 % Nonlinear vector field
3 f = [omega1;omega2;
4       M\ ( Bin*tau-diag ([b1 b2]) * [omega1;omega2]-C*[omega1;omega2]-grav ) ];
5 % Linearized model
6 Bsym = diff(f,tau);
7 Asym = [ diff(f,theta1) diff(f,theta2) diff(f,omega1) diff(f,omega2) ];
8 % Use the command 'subs' to insert values for parameters and state
```

## Introduction

## Inverted Pendulum

Modelling

Linearization

Stability Analysis

Controllability Analysis

Stabilization using State Feedback

## Acrobot

Modelling

Linearization

Controllability Analysis

Stabilization using State Feedback

## Stability Analysis

Global Stability Analysis

Local Stability Analysis

# Controllability

## Controllability and State Feedback



The controllability of the Acrobot is determined by computing the controllability matrix  
( $n = 4$ )

$$\mathcal{C} = [B \quad AB \quad A^2B \quad A^3B]$$



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The linearized system is controllable; thus, we can design a state feedback controller by pole placement.

### Introduction

### Inverted Pendulum

Modelling

Linearization

Stability Analysis

Controllability Analysis

Stabilization using State Feedback

### Acrobot

Modelling

Linearization

Controllability Analysis

Stabilization using State Feedback

### Stability Analysis

Global Stability Analysis

Local Stability Analysis

# Stabilization using State Feedback

State Feedback of Linearized System



To stabilize the acrobot in upright, a feedback controller needs to be designed; it is possible since the linearized system is controllable. Poles must be placed in the open left-half plane, and the system will respond faster when the poles are more negative.

# Stabilization using State Feedback

## State Feedback of Linearized System



To stabilize the acrobot in upright, a feedback controller needs to be designed; it is possible since the linearized system is controllable. Poles must be placed in the open left-half plane, and the system will respond faster when the poles are more negative.

A controller for the pendulum with poles  $s = -1$ ,  $s = -2$ ,  $s = -3$ ,  $s = -4$  is

$$u = Fx$$

where

$$F = [220.34 \quad 53.54 \quad 92.45 \quad 31.54]$$



## Introduction

## Inverted Pendulum

Modelling

Linearization

Stability Analysis

Controllability Analysis

Stabilization using State Feedback

## Acrobot

Modelling

Linearization

Controllability Analysis

Stabilization using State Feedback

## Stability Analysis

Global Stability Analysis

Local Stability Analysis



# Global Stability Analysis

Definition: Stability



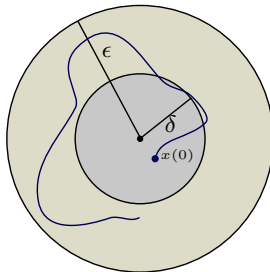
The equilibrium  $x = 0$  of a dynamical system

$$\dot{x} = f(x)$$

is

► *stable* if, for each  $\epsilon > 0$ , there is  $\delta$  such that

$$\|x(0)\| < \delta \Rightarrow \|x(t)\| < \epsilon, \quad \forall t \geq 0.$$



# Global Stability Analysis

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- *unstable* if it is not stable.
- *asymptotically stable* if it is stable and  $\delta$  can be chosen such that

$$\|x(0)\| < \delta \Rightarrow \lim_{t \rightarrow \infty} x(t) = 0.$$

# Global Stability Analysis

Lyapunov Stability



Let  $x = 0$  be an equilibrium point of  $\dot{x} = f(x)$ , and let  $V : \mathbb{R}^n \rightarrow \mathbb{R}$  be a differentiable. The equilibrium point  $x = 0$  is

► *stable* if  $V$  satisfies

$$V(0) = 0 \tag{1a}$$

$$V(x) > 0 \quad \forall x \in \mathbb{R}^n \setminus \{0\} \tag{1b}$$

$$\frac{\partial V}{\partial x}(x)f(x) \leq 0 \quad \forall x \in \mathbb{R}^n. \tag{1c}$$



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If the function  $V$  satisfies the above conditions then it is called a ***Lyapunov function***.

# Global Stability Analysis

Lyapunov Stability - Example



Consider the one-dimensional system

$$\dot{x} = -x^3$$

Determine if the system is globally asymptotically stable.

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Lyapunov Stability - Example



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It is seen that  $V(0) = 0$  and  $V(x)$  is positive for  $x \neq 0$ .

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This implies that  $V$  is a Lyapunov function and the system is globally asymptotically stable.

# Local Stability Analysis



## Introduction

## Inverted Pendulum

Modelling

Linearization

Stability Analysis

Controllability Analysis

Stabilization using State Feedback

## Acrobot

Modelling

Linearization

Controllability Analysis

Stabilization using State Feedback

## Stability Analysis

Global Stability Analysis

Local Stability Analysis

# Local Stability Analysis

## Local Lyapunov Stability



Let  $D \subset \mathbb{R}^n$ , let  $x = 0 \in D$  be an equilibrium point of  $\dot{x} = f(x)$ , and let  $V : D \rightarrow \mathbb{R}$  be a differentiable. The equilibrium point  $x = 0$  is *locally*

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Any region  $\Omega_\beta = \{x \in \mathbb{R}^n | V(x) \leq \beta\}$  such that  $\Omega_\beta \subseteq D$  is a positively invariant set contained in the region of attraction of the equilibrium point.



A set  $\Omega \subset \mathbb{R}^n$  is said to be an *invariant set* with respect to

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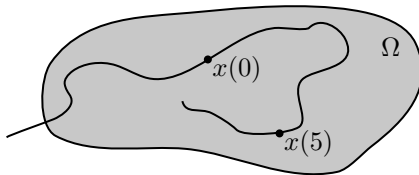
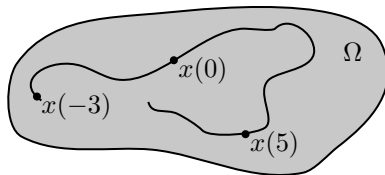
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# Local Stability Analysis

## Invariant Set

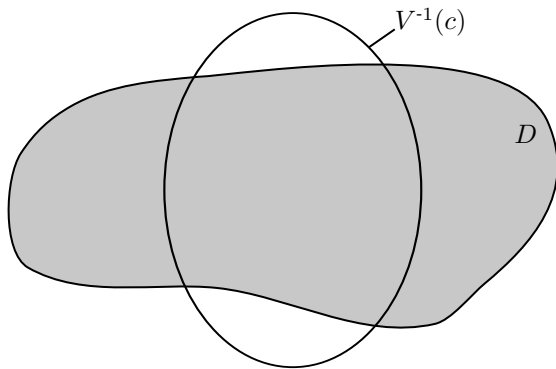


# Local Stability Analysis

Stability Region



Why do we say: "Any region  $\Omega_\beta = \{x \in \mathbb{R}^n | V(x) \leq \beta\}$  such that  $\Omega_\beta \subseteq D$  is a positively invariant set contained in the region of attraction of the equilibrium point."

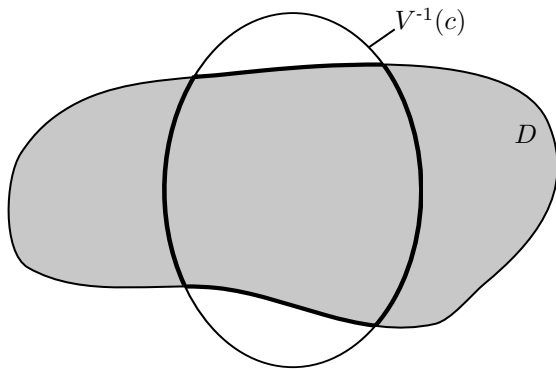


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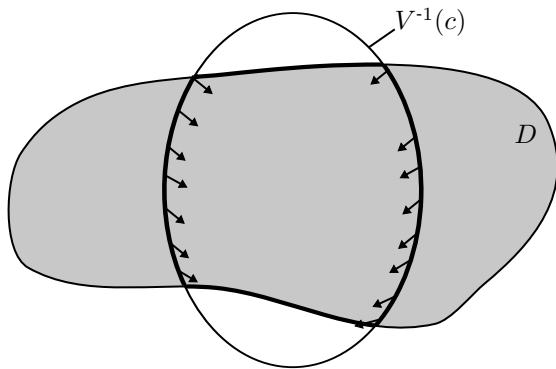


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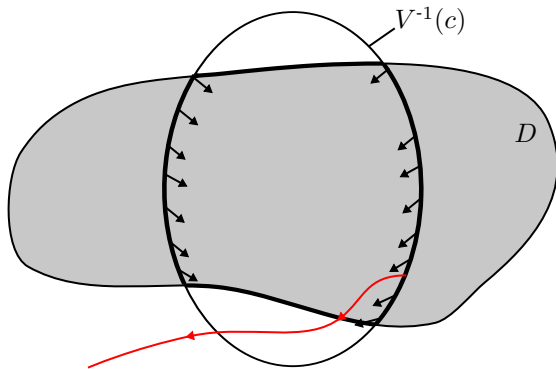


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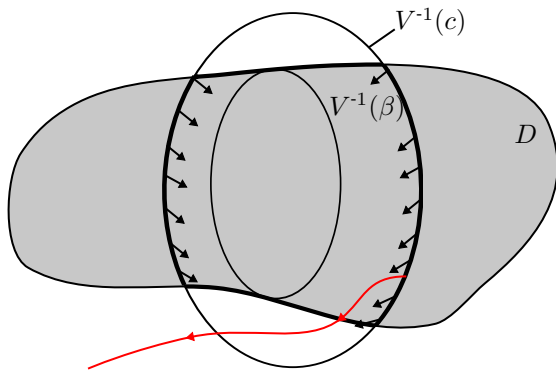


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