CIDBB Discussion 3 Walkthrough Stability Nonlinear Systems Control

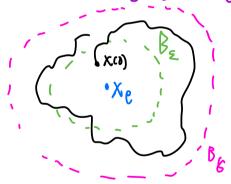
$$\dot{\chi} = f(x, \alpha)$$

x(t) goes to x_t as $t \to \infty$ Stability

XCH tracks XdCt)

Controllability





Stable in the sense of Lyapunav

·Xw) ~ Re

Asymptotic Stability

Any XLO) => blobal Asymptotic Stability
Within Bz => Local Asymptotic Stubility

Exponential Stability

[X(t) - Xel = eat | X(0) - Xel

X=Ax, max(Re(o(A))) <0 Linear System => Exponential Stubility => Exponential Stubility

X= \widetilde{\text{X}} \text{Re(\text{\sigma}(\text{\text{\$\infty}})=0}

Caxis of rotation

Stability in the Sease of Loggenor

C106B - Robotic Manipulation and Interaction

(Week 3)

Discussion #3

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Problem 1 - Lyapunov's Indirect Method: Modified Van Der Pol Oscillator

Consider the following model for an oscillator with nonlinear damping.

$$\ddot{x} + \mu(1 - x^2)\dot{x} + x = 0 \tag{0.1}$$

where μ is a scalar damping coefficient.

1. By choosing a good set of state variables, write the above model in state space form.

$$\begin{array}{ll}
\dot{x} = x_2 & \left[\dot{x}_1 \right] = f\left(\left[x_1 \right] \right) = \left[f(x_1, x_2) \right] \\
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2. Find all equilibria of this system.

$$\begin{bmatrix} 0 \\ 0 \end{bmatrix} = \begin{bmatrix} x_2 \\ -x_1 & -x_1(1-x_1^2)x_2 \end{bmatrix} \qquad \begin{cases} x_2 = 0 \\ -x_1 = 0 = 7x_1 = 0 \end{cases}$$

$$X_2 = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

3. Linearize the system about the equilibria. Using the indirect method of Lyapunov, comment on the stability of the equilibria for the cases where $\mu > 0$ and $\mu = 0$.

$$\begin{bmatrix} \dot{X}_{1} \\ \dot{X}_{2} \end{bmatrix} = \begin{bmatrix} f_{1} (x_{1}, x_{2}) \\ f_{2} (x_{1}, x_{2}) \end{bmatrix} = \begin{bmatrix} \dot{X}_{1} \\ \dot{X}_{2} \end{bmatrix} \begin{bmatrix} \dot{X}_{2} \\ \dot{X}_{1} \end{bmatrix} \begin{bmatrix} \dot{X}_{2} \\ \dot{X}_{2} \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ -1 & -M \end{bmatrix}$$

$$\lambda = -M + \sqrt{M^{2} - 4} \quad \text{Modified Asymptotic Stability}$$

$$\lambda \begin{bmatrix} \dot{X}_{1} \\ \dot{X}_{2} \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ -1 & -M \end{bmatrix}$$

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$$\lambda \begin{bmatrix} \dot{X}_{1} \\ \dot{X}_{2} \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ -1$$

- Discussion #3

Problem 2 - Lyapunov's Direct Method: Unicycle Model Robot

Consider the following model for a unicycle model robot. The state is (x, y, θ) which represents the position of the center of the robot relative to some fixed origin along with its current heading. The control inputs are the linear velocity v and the angular velocity ω .

$$\begin{bmatrix} \dot{x} \\ \dot{y} \\ \dot{\theta} \end{bmatrix} = \begin{bmatrix} v \cos \theta \\ v \sin \theta \\ \omega \end{bmatrix}$$
 (0.2)

In this problem, we will explore a technique called *point-offset* control for controlling unicycle model robots like the Turtlebot. Consider a point p attached rigidly to the robot at a distance δ from the center, in front of the robot (see figure 0.7). So, the position of p is given by:

$$\begin{bmatrix} p_x \\ p_y \end{bmatrix} = \begin{bmatrix} x + \delta \cos \theta \\ y + \delta \sin \theta \end{bmatrix} \tag{0.3}$$

Now consider the problem of driving the turtle bot to some neighbourhood of the origin. Instead of driving the turtle bot directly, we will instead attempt to control the robot so that the point p goes to the origin. Then, the turtle bot will be in a neighbourhood of radius δ around the origin. In the next few problems, we will develop a control law to drive p to the origin, and prove its stability.

1. Let the body frame axes of the turtlebot be $b_x = (\cos \theta, \sin \theta)^T$ and $b_y = (-\sin \theta, \cos \theta)^T$, as shown in figure 0.7. Show that

2. Say we apply the following feedback control law on the robot:

$$v = -b_x^T p, \qquad \omega = -\frac{1}{\delta} b_y^T p \tag{0.5}$$

Using the Lyapunov function

$$V = \frac{1}{2}p^T p - \frac{1}{2} \qquad (0.6)$$

show that the point p converges asymptotically to the origin. Is the stability global?

$$V = \frac{1}{2} \rho^{T} \rho , \quad \dot{V} = \rho^{T} \dot{\rho}$$

$$= \rho^{T} \dot{\rho} \quad \dot{V}(\rho) \angle O(\omega k s s \rho c o)$$

$$= \rho^{T} (V b_{x} + \omega k b_{y}) \quad \dot{U}(\rho) \angle O(\omega k s s \rho c o)$$

$$= \rho^{T} ((-6x^{T} \rho) b_{x} b_{x} (-\frac{1}{8} \cdot b_{y}^{T} \rho) b_{y}) \quad \dot{U}(\rho) b_{y} b_{y}$$

$$= -\rho^{T} \left[(b_{y} b_{y}^{T} \rho) b_{x} + (b_{y} b_{y}^{T} \rho) b_{y} \right]$$

$$= -\rho^{T} \left[b_{x} b_{y} \right] \left[\frac{6x^{T}}{6y^{T}} \right] \rho = -\rho^{T} \rho$$

$$= -V \rho U^{2}$$

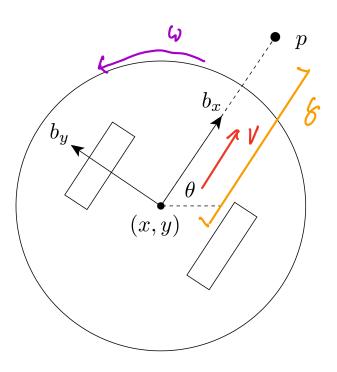
- Discussion #3

3. Is it exponentially stable? If so, is the stability global?

$$\dot{V} = -V \rho V^2$$

$$V = -2 V \qquad \text{Olobal Exponential}$$

$$V = \frac{1}{5} V \rho V^2 \qquad V(t) = e^{-2t} V(0)$$



(0.7)

X(t): f(x(t))

As t>00, X(x) > Xe Stablility

 $V(\chi): V(\chi) \geq 0, \quad V(\chi) = 0 \iff \chi = \chi_e$ "Charge function" $V(\chi): V(\chi) \geq 0, \quad V(\chi) = 0 \iff \chi = \chi_e$

· V(x) =0 (=> X= Xe