

Inverted Pendulum

Kishan Patel

December 11, 2024

1 Introduction

2 Lagrange Mechanics

$$\frac{d}{dt} \frac{\partial \mathcal{L}}{\partial \dot{q}_i} - \frac{\partial \mathcal{L}}{\partial q_i} = F_i \text{ with } \mathcal{L} = T - V \quad (1)$$

2.1 Finding the Lagrangian

$$\begin{aligned} T &= T_{\text{translational}} + T_{\text{rotational}} \\ &= T_c + T_w + T_{w_r} \\ &= \frac{1}{2} m_c v_c^2 + \frac{1}{2} m_w v_w^2 + \frac{1}{2} I_w \omega^2 \end{aligned} \quad (2)$$

The velocity vectors are in reference to the center of gravity, I will assume that the density of each component is constant. Hence, the rod's center of gravity is at $\frac{1}{2}l$ and the weight's, at the end of the rod, center of mass is at l .

$$\begin{aligned} v_c &= \begin{bmatrix} \dot{x} \\ 0 \end{bmatrix} \\ v_w &= \begin{bmatrix} \dot{x} + l\dot{\theta} \cos \theta \\ -l\dot{\theta} \sin \theta \end{bmatrix} \end{aligned} \quad (3)$$

Velocity squared is $v^2 = v \cdot v$

$$\begin{aligned} v_c^2 &= \dot{x}^2 \\ v_w^2 &= \dot{x}^2 + 2l\dot{x}\dot{\theta} \cos \theta + l^2\dot{\theta}^2 \cos^2 \theta + l^2\dot{\theta}^2 \sin^2 \theta \\ &= \dot{x}^2 + 2l\dot{x}\dot{\theta} \cos \theta + l^2\dot{\theta}^2 \end{aligned} \quad (4)$$

Now, simplify kinetic energy: T

$$\begin{aligned} T &= \frac{1}{2} m_c \dot{x}^2 \\ &+ \frac{1}{2} m_w \dot{x}^2 + m_w l \dot{x} \dot{\theta} \cos \theta + \frac{1}{2} m_w l^2 \dot{\theta}^2 \\ &+ \frac{1}{2} I_w \dot{\theta}^2 \\ T &= \frac{1}{2} \dot{x}^2 [m_c + m_w] + l \dot{x} \dot{\theta} \cos \theta m_w + \frac{1}{2} l^2 \dot{\theta}^2 [m_w + \frac{I_w}{l^2}] \end{aligned} \quad (5)$$

Now, will solve for the potential energy: V.

$$\begin{aligned} V &= V_w \\ &= m_w g l (1 - \cos \theta) \end{aligned} \quad (6)$$

Now, we can define the Lagrangian and solve for the mechanics with $q_1 = x$, and $q_2 = \theta$.

$$\begin{aligned}\mathcal{L} &= \frac{1}{2}\dot{x}^2[m_c + m_w] + \frac{1}{2}l^2\dot{\theta}^2[m_w + \frac{I_w}{l^2}] + l\dot{x}\dot{\theta}\cos(\theta)m_w - m_w gl(1 - \cos(\theta)) \\ &= \frac{1}{2}\dot{x}^2[m_c + m_w] + l^2\dot{\theta}^2 m_w + l\dot{x}\dot{\theta}\cos(\theta)m_w + m_w gl\cos(\theta) - m_w gl\end{aligned}\quad (7)$$

$q_1 = x$:

$$\begin{aligned}\frac{d}{dt}\frac{\partial\mathcal{L}}{\partial\dot{x}} &= \frac{\partial\mathcal{L}}{\partial x} + F_x \\ \frac{d}{dt}(\dot{x}[m_c + m_w] + l m_w \dot{\theta} \cos(\theta)) &= F_x \\ \ddot{x}[m_c + m_w] + l m_w \ddot{\theta} \cos(\theta) - l m_w \dot{\theta}^2 \sin(\theta) &= F_x\end{aligned}\quad (8)$$

$q_1 = \theta$:

$$\begin{aligned}\frac{d}{dt}\frac{\partial\mathcal{L}}{\partial\dot{\theta}} &= \frac{\partial\mathcal{L}}{\partial\theta} \\ \frac{d}{dt}(2 m_w \dot{\theta} l^2 + m_w \dot{x} \cos(\theta) l) &= -g l m_w \sin(\theta) - l m_w \dot{x} \sin(\theta) \\ 2 m_w \ddot{\theta} l^2 + m_w \ddot{x} \cos(\theta) l - m_w \dot{x} \dot{\theta} \sin(\theta) l &= -g l m_w \sin(\theta) - l m_w \dot{x} \sin(\theta) \\ 2 m_w \ddot{\theta} l^2 + m_w \ddot{x} \cos(\theta) l &= -g l m_w \sin(\theta)\end{aligned}\quad (9)$$

Since, the pendulum weight is being modeled as point mass $\frac{I_w}{l^2} = m_w$. sub in \ddot{x} into $\ddot{\theta}$. Now, sub $\ddot{\theta}$ back into \ddot{x} Using Matlab syms to simplify even more.

$$\begin{aligned}\ddot{x} &= \frac{2 l m_w \sin(\theta) \dot{\theta}^2 + 2 F_x + g m_w \cos(\theta) \sin(\theta)}{-m_w \cos(\theta)^2 + 2 m_c + 2 m_w} \\ \ddot{\theta} &= -\frac{l m_w \cos(\theta) \sin(\theta) \dot{\theta}^2 + F_x \cos(\theta) + g m_c \sin(\theta) + g m_w \sin(\theta)}{l (-m_w \cos(\theta)^2 + 2 m_c + 2 m_w)}\end{aligned}\quad (10)$$

To make the system more accurate, add friction terms $-\delta_c \dot{x}$ and $-\delta_p \dot{\theta}$ to \ddot{x} and $\ddot{\theta}$ respectively.

3 State Space Formulation

$$\text{Let } \mathbf{x} = \begin{bmatrix} x \\ \dot{x} \\ \theta \\ \dot{\theta} \end{bmatrix} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix}, \text{ and } F_x = u \quad (11)$$

$$\dot{\mathbf{x}} = \begin{bmatrix} \dot{x} \\ \ddot{x} \\ \dot{\theta} \\ \ddot{\theta} \end{bmatrix} = \begin{bmatrix} x_2 \\ \frac{2 l m_w \sin(x_3) x_4^2 + 2 u + g m_w \cos(x_3) \sin(x_3)}{-m_w \cos(x_3)^2 + 2 m_c + 2 m_w} \\ x_4 \\ -\frac{l m_w \cos(x_3) \sin(x_3) x_4^2 + u \cos(x_3) + g m_c \sin(x_3) + g m_w \sin(x_3)}{l (-m_w \cos(x_3)^2 + 2 m_c + 2 m_w)} \end{bmatrix} = f(x, u) \quad (12)$$

3.1 linearization about fixed point

Now, find states to linearized at $f(x, u) = 0$

$$f(x, u) = 0$$

$$f([x_1, 0, x_3, 0]^T, 0) = \begin{bmatrix} 0 \\ \frac{g m_w \cos(x_3) \sin(x_3)}{-m_w \cos(x_3)^2 + 2 m_c + 2 m_w} \\ 0 \\ -\frac{g m_c \sin(x_3) + g m_w \sin(x_3)}{l (-m_w \cos(x_3)^2 + 2 m_c + 2 m_w)} \end{bmatrix} \quad (13)$$

$f(x, u) = 0$ when $x_1 \in R$, $x_2 = 0$, $x_3 = \pi n \forall n \in Z$, $x_4 = 0$, and $u = 0$.

$$\bar{x} = \begin{bmatrix} x_1 \\ 0 \\ \pi n = 0 \\ 0 \end{bmatrix} \quad (14)$$

Now, we are ready for the state space equation $\dot{x} = Ax + Bu$. To find the matrices A and B we will take the Jacobian of $f(x, u)$. For the model, we will want x_3 to be 0 as that is the upright position of the pendulum.

$$\dot{x} = \partial_x f \Big|_{[x_1, 0, 0, 0]^T, 0} (x - \bar{x}) + \partial_u f \Big|_{[x_1, 0, 0, 0]^T, 0} (u - 0)$$

$$\dot{x} = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & -\delta & \frac{g m_w}{2 m_c + m_w} & 0 \\ 0 & 0 & 0 & 1 \\ 0 & -\delta & -\frac{g m_c + g m_w}{l (2 m_c + m_w)} & 0 \end{bmatrix} (x - \bar{x}) + \begin{bmatrix} 0 \\ \frac{2}{2 m_c + m_w} \\ 0 \\ -\frac{1}{l (2 m_c + m_w)} \end{bmatrix} u \quad (15)$$

$$y = Cx$$

Now, we have our state space equation and measurement equation. The δ term is friction of the cart moving horizontally.

3.2 Controllability

The rank($[B \ AB \ A^2B \ A^3B]$) = 4; therefore, we have full controllability of the system with our control $F_x = u$.

3.3 Observability

$$C = [1 \ 0 \ 0 \ 0] \quad (16)$$

To actualize the system, I will take the simplest case where the sensors will capture the positional state $x_1 = x$ only. Now calculate the rank Observability matrix to see if the system is full state observable from just the position of the cart.

$$O = [C^T \ A^T C^T \ (A^T)^2 C^T \ (A^T)^3 C^T] \quad (17)$$

The rank of the Observability matrix is 4; therefore, we can estimate all the other states.

4 Controller with $C = I$

4.1 Proportional Control

For the proportional control let $u - \bar{u} = -K(x - \bar{x})$.

$$\begin{aligned} \dot{x} &= Ax + Bu \\ \dot{x} &= Ax - BKx \\ \dot{x} &= (A - BK)x \end{aligned} \quad (18)$$

Since the system is controllable, the matrix K can be placed so that $A - BK$ can have any eigenvalues. Now what eigenvalues should one choose? $\lambda_i \in \{z \in C : \text{Re}(z) < 0\}$. But, what is the optimal choice? For this, the Linear Quadratic Regulator (L.Q.R.) will be used. Define the cost J which is a quadratic function in x and u .

$$\begin{aligned}
 J &= \int_0^\infty x^T Q x + u^T R u \, dt \\
 \dot{x} &= Ax + Bu \\
 Q &= \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 10 & 0 & 0 \\ 0 & 0 & 10 & 0 \\ 0 & 0 & 0 & 100 \end{bmatrix} \\
 R &= [1]
 \end{aligned} \tag{19}$$

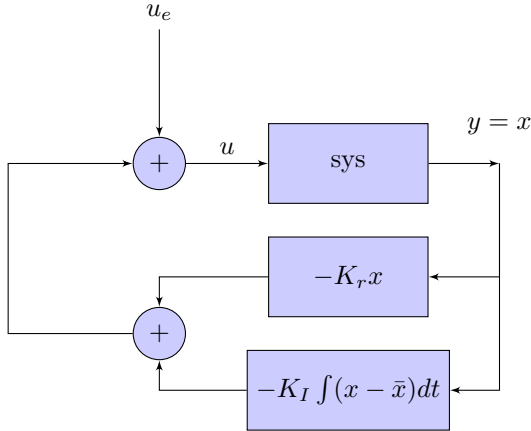
Now finding the minimum of the cost function is all done for us in Matlab. $K_r = \text{lqr}(A, B, Q, R)$. The interpretation of Q and R is that it defines the cost of controlling specific states.

4.2 Integral Control

For the proportional control let $u - \bar{u} = -K_I \int_0^t (x - \bar{x}) dt$.

$$\begin{aligned}
 w &= \int_0^t (x - \bar{x}) dt \\
 \dot{w} &= x - \bar{x} \\
 w(0) &= 0
 \end{aligned} \tag{20}$$

To implement integral action set up a new state w which will update with the following discrete rule $w_n = w_{n-1} + \dot{w}_n dt$. The internal gain $K_I = [.1, .1, .1, .1]$.



5 Controller with $C = [1 \ 0 \ 0 \ 0]$

Since the controller has access to only the positional state, an estimator must be built to estimate other states.

5.1 Estimator

A system needs to build that will give us \hat{x} . The estimate \hat{x} needs to approach the actual state x . So the error $e = x - \hat{x}$ needs to approach 0.

$$\begin{aligned}
e &= x - \hat{x} \\
\dot{e} &= \dot{x} - \dot{\hat{x}} \\
\dot{e} &= Ax + Bu - (A\hat{x} + Bu + f) \\
\dot{e} &= A(x - \hat{x}) - L(y - \hat{y}) \\
\dot{e} &= (A - LC)e
\end{aligned} \tag{21}$$

So if L can be placed such that $A - LC$ has negative eigenvalues, then the estimator will approach the actual state, that is, if the observability matrix has full rank.

$$\begin{aligned}
\dot{\hat{x}} &= A\hat{x} + Bu + L(y - \hat{y}) \\
\hat{y} &= C\hat{x}
\end{aligned} \tag{22}$$

5.2 Kalman Filter

The matrix L needs to be chosen optimally to make the estimator. Let V_d be the covariance matrix for the disturbance in the system, and let L_n be the covariance matrix for the noise in the measurement.

$$\begin{aligned}
\dot{x} &= Ax + Bu + w_d \\
y &= Cx + w_n
\end{aligned} \tag{23}$$

$$K_f = lqr(A', C', V_d, V_n)' \tag{24}$$

