

ORF 418 : OPTIMAL LEARNING

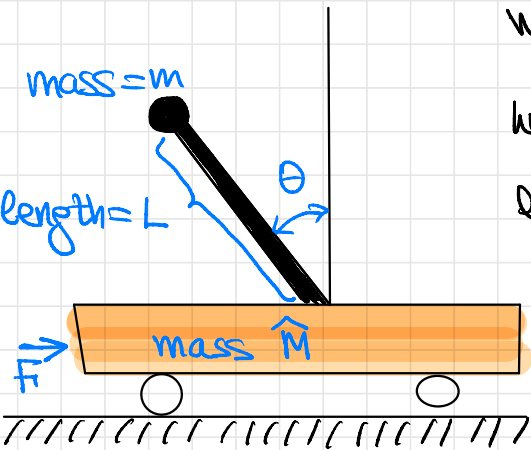
LECTURE 5 : September 17, 2025

Inverted Pendulum

1. Formulation
2. Controllability
3. Connection to Riccati Equations and LQR.



I. INVERTED PENDULUM



We want to keep an inverted pendulum on a cart close to the vertical line (i.e., $\theta=0$) by moving the cart horizontally. Clearly this is an **unstable** configuration.

a) Problem formulation. State variables are

$\theta(t)$ = angle, $\dot{\theta}(t)$ = derivative of the angle

$p(t)$ = position of the cart, $p'(t)$ = cart velocity.

So we set

$$x(t) = (\theta(t), \dot{\theta}(t), p(t), p'(t))^T.$$

Control variable is the force $F(t)$ applied to the cart.

Dynamics come from classical mechanics:

$$F(t) = (\hat{M} + m)p''(t) - mL\theta''(t)\cos\theta(t) + mL|\dot{\theta}(t)|^2\sin\theta(t)$$

$$L\theta''(t) = g\sin\theta(t) + p''(t)\cos\theta(t),$$

where g is the gravitational constant $\approx 9.81 \approx 10 \text{ m/sec}^2$

Note that $x(t) = (0, 0, 0, 0)^T$ is an unstable solution.

Difficulties are several:

(i) dynamics are not linear.

(ii) it is naturally set in continuous time.

(iii) objective function is not specified (yet).

(b) Linearization Since the pendulum is going to operate $\theta(t) \approx 0$, we linearize the above functions around the $\theta=0$ by using Taylor's formula:

$$\cos\theta \approx 1 - \theta^2 \approx 1, \quad \sin\theta \approx \theta, \quad |\theta'(t)|^2 \approx 0.$$

This yields the simplified equations

$$F(t) = (\hat{M} + m) p''(t) - m l \theta''(t),$$

$$l\theta(t)'' = g\theta(t) + p''(t).$$

Let $x(t)$ be as above and set

$$u(t) := \frac{F(t)}{L\hat{M}}$$

Then,

$$\frac{d}{dt} x(t) = \frac{d}{dt} \begin{bmatrix} p(t) \\ p'(t) \\ \theta(t) \\ \theta'(t) \end{bmatrix} = \underbrace{\begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & a & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & b & 0 \end{bmatrix}}_{= A_c} x(t) + \underbrace{\begin{bmatrix} 0 \\ L \\ 0 \\ 1 \end{bmatrix}}_{= B_c} u(t)$$

where $a = mg/\hat{m}$, $b = (m + \hat{M})g/e\hat{m}$.

Not finished yet! We still need to discretize time.

Choose a small time step $h = \Delta t$. Then,

$$\frac{1}{h} [x(t+h) - x(t)] \approx \frac{d}{dt} x(t) = A_c x(t) + B_c u(t)$$

By algebra we obtain

$$x(t+h) \approx x(t) + h A_c x(t) + h B_c u(t)$$

$$= \underbrace{[I + h A_c]}_A x(t) + \underbrace{h B_c}_B u(t)$$

Hence

$$A = I + A_c = \begin{bmatrix} 1 & h & 0 & 0 \\ 0 & 1 & ah & 0 \\ 0 & 0 & 1 & h \\ 0 & 0 & bh & 1 \end{bmatrix}, \quad B = h B_c = \begin{bmatrix} 0 \\ hL \\ 0 \\ h \end{bmatrix}$$

This establishes a linear equation for the state process in discrete time.

For LQ control, we need a quadratic cost function or equivalently matrices M and N . But the problem is a question of controllability:

is there a control that bring the state to zero?

II. CONTROLLABILITY

We say that the system in \mathbb{R}^d :

$$x_{k+1} = Ax_k + Bu_k \quad (*)$$

with given matrices A, B is **controllable** if for every initial point x_{in} and a terminal point x_{out} there are $n \geq 1$ and $u = (u_0, \dots, u_n) \in \mathcal{U}$ such that the solution to $(*)$ starting at $x_0 = x_{in}$ satisfies $x_n = x_{out}$. In words, by choosing appropriate controls we can move the state from any point x_{in} to any other point x_{out} in finitely many steps.

THEOREM 2.3.1 The dynamical system $(*)$ is controllable if and only if the $d \times d$ controllability matrix

$$P = [B \mid AB \mid \dots \mid A^{d-1}B]$$

has (full) rank d . Moreover, when the system is controllable then the Riccati equation with any M and N has a unique solution.

See the notes for a proof.

Pendulum Example. We have $d=4$ $l=1$. Then,

$$A = \begin{bmatrix} 1 & h & 0 & 0 \\ 0 & 1 & ah & 0 \\ 0 & 0 & 1 & h \\ 0 & 0 & bh & 1 \end{bmatrix}, \quad B = \begin{bmatrix} 0 \\ hL \\ 0 \\ h \end{bmatrix}$$

We compute that

$$AB = \begin{bmatrix} h^2 L \\ hL \\ h^2 \\ h \end{bmatrix} \quad A^2 B = \begin{bmatrix} 2h^2 L \\ hL + ah^3 \\ 2h^3 \\ bh^3 + h \end{bmatrix} \quad A^3 B = \begin{bmatrix} 3h^2 L + ah^4 \\ -hL + 3ah^3 \\ 3h^2 + bh^4 \\ 3bh^3 + h \end{bmatrix}$$

$$\mathcal{O} = h \begin{bmatrix} 0 & Lh & 2Lh & 3Lh + ah^3 \\ L & L & L + ah^2 & L + 3ah^2 \\ 0 & h & 2h^2 & 3h + bh^3 \\ 1 & 1 & bh^2 + 1 & 3bh^2 + 1 \end{bmatrix}$$

has full rank (has to be checked).

III. Connection to LQR

THEOREM 2.3.2. Suppose that the dynamical system with matrices A and B is controllable.

Then, for any positive semi-definite, symmetric matrices M and N , the Riccati equation is solvable.

Proof. For $n \geq 1$ consider

$$\begin{aligned} v_n(x_0) &:= \inf_u J_n(x_0, u) \\ &= \inf_u \sum_{k=0}^{n-1} e^{kR} [x_k^T M x_k + u_k^T N u_k] \end{aligned}$$

And

$$\begin{aligned} v_\infty(x_0) &= \inf_u J_\infty(x_0, u) \\ &= \inf_u \sum_{k=0}^{\infty} e^{kR} [x_k^T M x_k + u_k^T N u_k] \end{aligned}$$

Since the system is controllable, there is $u \in \mathcal{U}$ such that $x_n = 0$. Hence,

$$v_\infty(x_0) < \infty \quad \text{and} \quad \lim_{n \rightarrow \infty} v_n(x_0) = v_\infty(x_0).$$

By induction we can show that

$$v_n(x_0) = x_0^T v_n x_0$$

for some positive definite symmetric matrix

v_n . By convergence of $v_n(x_0)$, we conclude that matrices v_n 's also converge. Hence,

$-V_0(x_0) = x_0^T V_0 x_0$ and we then show that V_0 solves the Riccati equation. \square

Inverted Pendulum Example.

Suppose $m=1$, $\hat{M}=100$, $L=1$ and take $g \approx 10$. Then, $a=0.1$ and $b=10.1$. We arbitrarily choose $h=\Delta t=0.1$, $\rho=0.36$, $M=I$ and $N=1$. Let V be the solution of the Riccati equation which we know to exist. Then the gain matrix

F is given by

$$F = (N + B^T V B)^{-1} B^T V A$$

Using the package `Enalg` from `spicy` in python and the command

`solve_discrete_are(...)`

we compute V . Here "are" stands for Algebraic Riccati Equations. We can then compute F . Results are in lecture notes.