

Kuwait University
College of Engineering and Petroleum



جامعة الكويت
KUWAIT UNIVERSITY

ME417 CONTROL OF MECHANICAL SYSTEMS

PART II: CONTROLLER DESIGN VIA STATE-SPACE

LECTURE 4: INTRODUCTION TO LINEAR OPTIMAL CONTROL

Summer 2020

Ali AlSaibie

- Objectives:
 - Introduce the concept of Optimal Control
 - Introduce the Linear Quadratic Regulator Problem LQR
- Reading:
 - *Modern Control Systems, Dorf & Bishop. Chapter 11 – Section on Optimal Control Systems*



- Suppose you are working on a construction project
- You want to optimize the outcomes of the project:
 - You wish to minimize time and money, and maximize quality.
- Assume you can only control the effort you input into the system. Stated mathematically:

$$\dot{\mathbf{x}} = \mathbf{f}(\mathbf{x}, u, t) = \mathbf{A}(\mathbf{x}, t) + \mathbf{B}(u, t)$$

$$\text{Where } \mathbf{x} = \left[\text{Time} \quad \text{Money} \quad \frac{1}{\text{Quality}} \right]^T, u = \text{Effort}$$

- Suppose for simplicity, that the system is **LTI** (absurd simplification)

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



- Suppose you apply a feedback controller

$$u_{effort} = \mathbf{K}(\mathbf{x}_{desired} - \mathbf{x})$$
$$\dot{\mathbf{x}} = (\mathbf{A} - \mathbf{BK})\mathbf{x} + \mathbf{BK}\mathbf{x}_{desired}$$

- $\mathbf{x}_{desired}$: Can be your desired target for money, time and quality for instance.
- You want to find the values of the gain matrix \mathbf{K} that optimizes your project outcome:
 - So that you apply just the right amount of effort to achieve your goal.
- To achieve your optimization goal, you first must come up with a cost function in which you weight the relative importance of each state and input

$$f_{cost} = \int_{t=0}^{t=\infty} w_1 \cdot Time + w_2 \cdot Money + w_3 \frac{1}{Quality} + w_4 \cdot Effort$$



$$f_{cost} = \int_{t=t_0}^{t=t_1} w_1 \cdot Time + w_2 \cdot Money + w_3 \frac{1}{Quality} + w_4 \cdot Effort$$

- The cost function computes the weighted sum of the four measures over a period of time.
- The coefficients w_1, \dots, w_4 are weights we, the problem designers, place in the cost function based on the relative importance we give to each measure.
- So, suppose you want to penalize delays in the project heavily but can tolerate additional expenses, then you put more weight w_1 relative to w_2
- The weights have to be scaled to account for the units/scale of each variable.
 - Is delaying one extra day as bad as spending one extra KWD?



The Optimization Problem - Conceptually

- The solution for the gain matrix \mathbf{K} , is computed by minimizing the cost function
$$\min f_{cost} \Rightarrow \partial f_{cost} = 0$$
Subject to: $\dot{\mathbf{x}} = (\mathbf{A} - \mathbf{BK})\mathbf{x} + \mathbf{BK}\mathbf{x}_{desired}$
- Plus any other constraints we put on the system/project
 - Such as maximum effort, maximum delay, minimum quality, etc.
- In the optimization problem above, what you tune are the weights you give in the cost function, while the gains in \mathbf{K} are to be *computed* by “mathematical” optimization.
- The cost function, is also referred to as the “performance index”



Optimal Controller - Example

- In applying the concept of optimization to our state-feedback control structure.
 - We don't "place" the poles of the system directly, but instead, solve the optimization problem, that would indirectly place the poles of the system for us, and consequently, compute the state feedback gain \mathbf{K}

- Given the following LTI mechanical system

$$\begin{aligned}\dot{\mathbf{x}} &= \begin{bmatrix} 0 & 1 \\ -1 & -1 \end{bmatrix} \mathbf{x} + \begin{bmatrix} 0 \\ 1 \end{bmatrix} u \\ \mathbf{y} &= \begin{bmatrix} 1 & 0 \end{bmatrix} \mathbf{x}\end{aligned}$$

- We wish to apply a state feedback controller that regulates the system, and equally minimizes the states x_1 and x_2 over time (position and velocity).
 - Regulates: returns the system to $\mathbf{x} = \mathbf{0}$, with $r = 0$
 - In the regulation problem the states correspond to the error $\mathbf{e} = \mathbf{0} - \mathbf{x}$



$$u = -\mathbf{K}\mathbf{x}$$
$$\dot{\mathbf{x}} = (\mathbf{A} - \mathbf{BK})\mathbf{x} = \tilde{\mathbf{A}}\mathbf{x}$$

- We set the cost function (performance index) as

$$J = \int_0^{\infty} (x_1^2 + x_2^2) dt$$

- Note that $x_1^2 + x_2^2 = [x_1 \quad x_2] \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \mathbf{x}^T \mathbf{x}$, so

$$J = \int_0^{\infty} (\mathbf{x}^T \mathbf{x}) dt$$

- To minimize the performance index J , we assume a form for J s.t.

$$\frac{d}{dt}(J) = \frac{d}{dt}(\mathbf{x}^T \mathbf{P} \mathbf{x}) = -\mathbf{x}^T \mathbf{x}$$



- If we determine the constant matrix \mathbf{P} , then we solve for $\frac{d}{dt}(\mathbf{x}^T \mathbf{P} \mathbf{x}) = 0$, to minimize the performance index

$$\frac{d}{dt}(\mathbf{x}^T \mathbf{P} \mathbf{x}) = \dot{\mathbf{x}}^T \mathbf{P} \mathbf{x} + \mathbf{x}^T \mathbf{P} \dot{\mathbf{x}}$$

- Substitute $\dot{\mathbf{x}} = \tilde{\mathbf{A}}\mathbf{x}$, in the above equation

$$\frac{d}{dt}(\mathbf{x}^T \mathbf{P} \mathbf{x}) = (\tilde{\mathbf{A}}\mathbf{x})^T \mathbf{P} \mathbf{x} + \mathbf{x}^T \mathbf{P}(\tilde{\mathbf{A}}\mathbf{x}) = \mathbf{x}^T \tilde{\mathbf{A}} \mathbf{P} \mathbf{x} + \mathbf{x}^T \mathbf{P} \tilde{\mathbf{A}} \mathbf{x} = \mathbf{x}^T (\tilde{\mathbf{A}} \mathbf{P} + \mathbf{P} \tilde{\mathbf{A}}) \mathbf{x}$$

- Using the matrix transpose property: $(\mathbf{A}\mathbf{B})^T = \mathbf{B}^T \mathbf{A}^T$

- Let's choose a \mathbf{P} such that $\tilde{\mathbf{A}} \mathbf{P} + \mathbf{P} \tilde{\mathbf{A}} = -\mathbf{I}$, then

$$\frac{d}{dt}(\mathbf{x}^T \mathbf{P} \mathbf{x}) = \mathbf{x}^T (-\mathbf{I}) \mathbf{x} = -\mathbf{x}^T \mathbf{x}, \text{ the same equation we formed earlier}$$



- Now we substitute and integrate the performance index

$$J = \int_0^{\infty} \frac{d}{dt} (J) dt = \int_0^{\infty} \frac{d}{dt} (\mathbf{x}^T \mathbf{P} \mathbf{x}) dt = -\mathbf{x}^T \mathbf{P} \mathbf{x} \Big|_0^{\infty} = \mathbf{x}^T(0) \mathbf{P} \mathbf{x}(0)$$

- Since we assume the system is stable and the states will return to $\mathbf{0}$, as $t \rightarrow \infty$
 $\mathbf{x}^T(\infty) \mathbf{P} \mathbf{x}(\infty) = 0$
- The solution to our optimal controller given the performance index we formed is to find a \mathbf{P} that satisfies the following equation

$$\tilde{\mathbf{A}} \mathbf{P} + \mathbf{P} \tilde{\mathbf{A}} = (\mathbf{A} - \mathbf{B} \mathbf{K}) \mathbf{P} + \mathbf{P} (\mathbf{A} - \mathbf{B} \mathbf{K}) = -\mathbf{I}$$

- The solution will be different if we choose a different performance index form.



- In the previous example we postulated the performance index such that it equally minimizes the position and velocity errors
- But what if we want to penalize the states with relatively different weights?
 - Rather than equally penalize both position and velocity errors, we may want to penalize position error more than velocity error: $w_1 > w_2$

$$J = \int_0^{\infty} (w_1 x_1^2 + w_2 x_2^2) dt = \int_0^{\infty} \mathbf{x}^T \begin{bmatrix} w_1 & 0 \\ 0 & w_2 \end{bmatrix} \mathbf{x} dt$$

- Perhaps we don't even care about velocity, we just want to get to position zero as fast as possible

$$J = \int_0^{\infty} x_1^2 dt = \int_0^{\infty} \mathbf{x}^T \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} \mathbf{x} dt$$



- And what if we want to consider the input u ?
 - Perhaps we have limited battery power and wasting a lot of energy to regulate the system is not desired.
 - We can include the input u as part of the performance index

$$J = \int_0^{\infty} (w_1 x_1^2 + w_2 x_2^2 + w_3 u^2) dt$$

- The general formulation for the above performance index is captured in what is called, the linear quadratic regulator, or LQR. Where the performance index is defined as

$$J = \int_0^{\infty} (q_{11}x_1^2 + \cdots + q_{nn}x_n^2) + (r_1u_1^2 + \cdots + r_mu_m^2) dt$$

- ignoring the coupling weights (e.g. $q_{12}x_1x_2$)



The Linear Quadratic Regulator

- Given a closed-loop LTI system with full state feedback: $u = r - \mathbf{K}x$

$$\dot{x} = (\mathbf{A} - \mathbf{BK})x + \mathbf{B}r$$
$$y = \mathbf{C}x$$

- But there is an optimal way to compute \mathbf{K} , the gain vector, via the Linear Quadratic Regulator (LQR)
- The LQR performance index

$$J = \int_0^{\infty} (x^T \mathbf{Q}x + u^T \mathbf{R}u) dt$$

\mathbf{Q} is a matrix we chose, to give weights to each state in x

\mathbf{R} is a weighting factor for the input to the system u



The Linear Quadratic Regulator

- The optimal \mathbf{K} in $u = r - \mathbf{K}\mathbf{x}$ is found by minimizing J , subject to the closed-loop system dynamics:

$$\begin{aligned} \min J &= \int_0^\infty (\mathbf{x}^T \mathbf{Q} \mathbf{x} + u^T \mathbf{R} u) dt \\ \text{s.t. } \dot{\mathbf{x}} &= (\mathbf{A} - \mathbf{B}\mathbf{K})\mathbf{x} + \mathbf{B}r \end{aligned}$$

- After some more math....
- The solution for \mathbf{K} is given by

$$\mathbf{K} = \mathbf{R}^{-1} \mathbf{B}^T \mathbf{P}$$

- Where \mathbf{P} is found from solving the Riccati Equation:

$$\mathbf{A}^T \mathbf{P} + \mathbf{P} \mathbf{A} - \mathbf{P} \mathbf{B} \mathbf{R}^{-1} \mathbf{B}^T \mathbf{P} + \mathbf{Q} = \mathbf{0}$$

- This is not meant to be solved by hand, and we can use MATLAB's `lqr()` function to solve for \mathbf{K} , directly.
 - The function `lqr()` also returns the optimal pole locations of the Closed-Loop system.
- \mathbf{R} and \mathbf{Q} must be positive definite



- As a starting point, choose $\mathbf{Q} = \mathbf{I}$ and $R = 1$
- To improve \mathbf{Q} , place more/less weights on the states you care more/less to stabilize (regulate: bring to zero).
- If you care to minimize the input to the system (the effort/energy you spend on the system), then you might want to increase R and vice versa.
- You can generally achieve a good outcome by keeping the \mathbf{Q} matrix diagonal
 - Unless you want to weight a coupling effect between states.

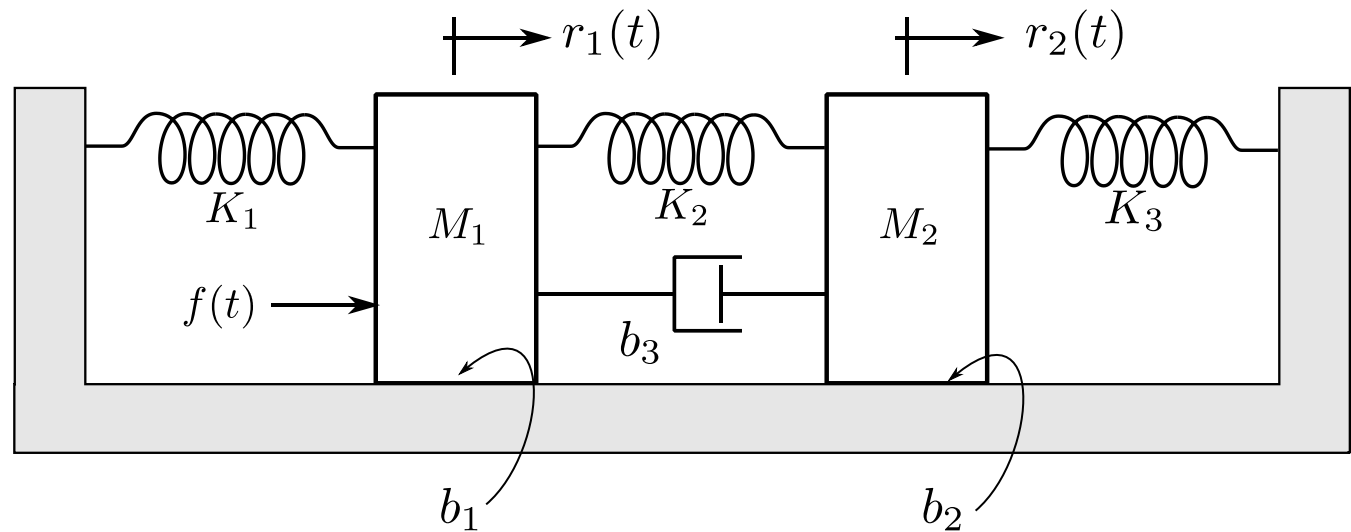
LQR MATLAB Example

- Given the two degree of freedom system shown, whose state-space representation is given by

$$\dot{\mathbf{x}} = \begin{bmatrix} 0 & 1 & 0 & 0 \\ \frac{-(K_1 + K_2)}{m_1} & \frac{-(b_1 + b_3)}{m_1} & \frac{K_2}{m_1} & \frac{b_3}{m_1} \\ 0 & 0 & 0 & 1 \\ \frac{K_2}{m_2} & \frac{b_3}{m_2} & \frac{-(K_2 + K_3)}{m_2} & \frac{-(b_2 + b_3)}{m_2} \end{bmatrix} \mathbf{x} + \begin{bmatrix} 0 \\ \frac{1}{m_1} \\ 0 \\ 0 \end{bmatrix} u$$

$$\mathbf{y} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \mathbf{x}$$

Let's design a feedback controller $u = r - \mathbf{K}\mathbf{x}$, to stabilize the system from a nonzero initial condition.



- See "*PIII_L4_Examples.m*"

