



# Chapter 6

## Optimal Control

# Outlines

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- **Basic concepts**
- **Problem formulation**
- **Linear quadratic optimal control**

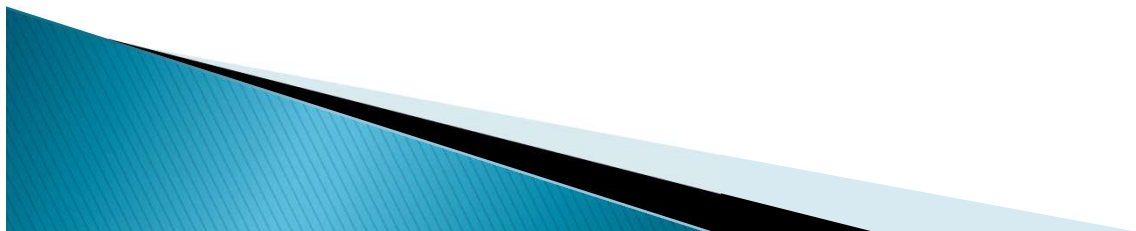


# Basic concepts

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**Optimal control** is the key concept in Modern Control Theory

Given state equations and certain constraints, try to find the optimal control law so that the system has optimal performance.



# Problem formulation

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## Four important factors

### (1) General form of state equation

$$\dot{x} = f(x(t), u(t), t), x(t_0) = x_0$$

### (2) Target set: The state can be some point set S.

$$S = \{x_{t_f} | \Psi[x(t_f)] = 0\}$$

### (3) Control u is bounded, that is $u \in \Omega$

### (4) Performance criteria:

$$J = \varphi[x(t_f)] + \int_{t_0}^{t_f} L[x(t), u(t), t] dt$$



Restriction on  
final state

Restriction on state and  
control

# Problem formulation

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$$J = \varphi[\mathbf{x}(t_f)] + \int_{t_0}^{t_f} L[\mathbf{x}(t), u(t), t] dt$$

(1) If  $\varphi[\mathbf{x}(t_f)] = 0$ , we have **Integral performance**

$$J = \int_{t_0}^{t_f} L[\mathbf{x}(t), u(t), t] dt$$

(2) If  $L[\mathbf{x}(t), u(t), t] = 0$ , we have **end performance**

$$J = \varphi[\mathbf{x}(t_f)]$$

(3) Otherwise, it has the general form as shown above.



# Linear quadratic optimal control

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Suppose the dynamic equation  $\dot{\mathbf{x}}(t) = \mathbf{A}(t)\mathbf{x}(t) + \mathbf{B}(t)\mathbf{u}(t)$   
 $\mathbf{y}(t) = \mathbf{C}(t)\mathbf{x}(t)$

where  $\mathbf{x}(t) \in \mathbb{R}^n, \mathbf{u}(t) \in \mathbb{R}^p, \mathbf{y}(t) \in \mathbb{R}^q$

The functional performance is

$$J = \frac{1}{2} \mathbf{x}^T(t_f) \mathbf{F} \mathbf{x}(t_f) + \frac{1}{2} \int_{t_0}^{t_f} [\mathbf{x}^T(t) \mathbf{Q} \mathbf{x}(t) + \mathbf{u}^T(t) \mathbf{R} \mathbf{u}(t)] dt$$

$$\mathbf{F} \in \mathbb{R}^{n \times n} \geq 0, \mathbf{F} = \mathbf{F}^T, \mathbf{Q} \in \mathbb{R}^{n \times n} \geq 0, \mathbf{Q} = \mathbf{Q}^T, \mathbf{R} \in \mathbb{R}^{r \times r} > 0, \mathbf{R} = \mathbf{R}^T,$$

Try to find  $\mathbf{u}(t) \Rightarrow J_{\min}$



# Linear quadratic optimal control

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$$J = \frac{1}{2} \mathbf{x}^T(t_f) \mathbf{F} \mathbf{x}(t_f) + \frac{1}{2} \int_{t_0}^{t_f} [\mathbf{x}^T(t) \mathbf{Q} \mathbf{x}(t) + \mathbf{u}^T(t) \mathbf{R} \mathbf{u}(t)] dt$$

The meanings of each term:

$\frac{1}{2} \mathbf{x}^T(t_f) \mathbf{F} \mathbf{x}(t_f)$  — It is related to final state, and each state is controlled by F

$\frac{1}{2} \mathbf{x}^T(t) \mathbf{Q} \mathbf{x}(t)$  — It is related to the process, and each state is controlled by Q

$\frac{1}{2} \mathbf{u}^T(t) \mathbf{R} \mathbf{u}(t)$  — It is related to control that minimize the energy consumption



# Linear quadratic optimal control

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**State regulator:**

$$J = \frac{1}{2} \mathbf{x}^T(t_f) \mathbf{F} \mathbf{x}(t_f) + \frac{1}{2} \int_{t_0}^{t_f} [\mathbf{x}^T(t) \mathbf{Q} \mathbf{x}(t) + \mathbf{u}^T(t) \mathbf{R} \mathbf{u}(t)] dt$$

**Output regulator:**

$$J = \frac{1}{2} \mathbf{y}^T(t_f) \mathbf{F} \mathbf{y}(t_f) + \frac{1}{2} \int_{t_0}^{t_f} [\mathbf{y}^T(t) \mathbf{Q} \mathbf{y}(t) + \mathbf{u}^T(t) \mathbf{R} \mathbf{u}(t)] dt$$

**Output tracking regulator**

$$J = \frac{1}{2} \mathbf{e}^T(t_f) \mathbf{F} \mathbf{e}(t_f) + \frac{1}{2} \int_{t_0}^{t_f} [\mathbf{e}^T(t) \mathbf{Q} \mathbf{e}(t) + \mathbf{u}^T(t) \mathbf{R} \mathbf{u}(t)] dt$$





# Linear quadratic optimal control

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## (1) When $t_f$ is fixed

**Linear time-varying state space model**

$$\begin{aligned}\dot{\mathbf{x}}(t) &= \mathbf{A}(t)\mathbf{x}(t) + \mathbf{B}(t)\mathbf{u}(t), & \mathbf{x}(t_0) &= \mathbf{x}_0 \\ y(t) &= \mathbf{C}(t)\mathbf{x}(t)\end{aligned}$$

**The functional performance**  $J = \frac{1}{2} \mathbf{x}^T(t_f) \mathbf{F} \mathbf{x}(t_f) + \frac{1}{2} \int_{t_0}^{t_f} [\mathbf{x}^T(t) \mathbf{Q} \mathbf{x}(t) + \mathbf{u}^T(t) \mathbf{R} \mathbf{u}(t)] dt$

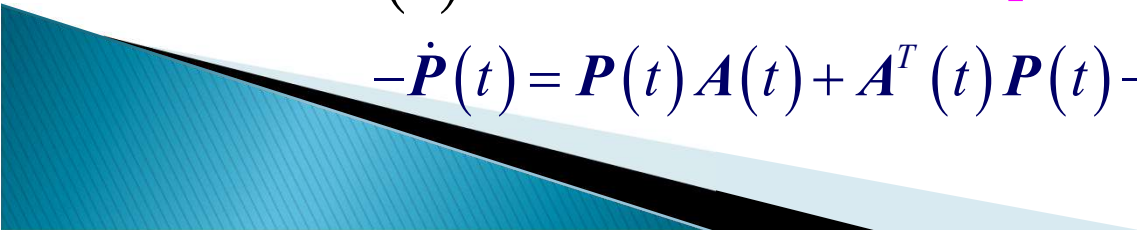
where  $\mathbf{Q}(t) \geq 0$ ,  $\mathbf{R}(t) > 0$ ,  $\mathbf{F} \geq 0$

Try to find  $\mathbf{u}^*$   $J \rightarrow \min$

There exists the optimal control law  $\mathbf{u}^*(t) = -\mathbf{R}^{-1}(t) \mathbf{B}^T(t) \mathbf{P}(t) \mathbf{x}(t)$

Optimal performance  $J^* = \frac{1}{2} \mathbf{x}^T(t_0) \mathbf{P}(t_0) \mathbf{x}(t_0)$

where  $\mathbf{P}(t)$  satisfies Ricatti equation

$$-\dot{\mathbf{P}}(t) = \mathbf{P}(t) \mathbf{A}(t) + \mathbf{A}^T(t) \mathbf{P}(t) - \mathbf{P}(t) \mathbf{B}(t) \mathbf{R}^{-1} \mathbf{B}^T(t) \mathbf{P}(t) + \mathbf{Q}(t)$$


# Linear quadratic optimal control

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Try to find  $\mathbf{u}^*$   $J \rightarrow \min$

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Optimal performance  $J^* = \frac{1}{2} \mathbf{x}^T(t_0) \mathbf{P}(t_0) \mathbf{x}(t_0)$

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Boundary condition  $\mathbf{P}(t_f) = \mathbf{F}$

Optimal trajectory  $\mathbf{x}^*(t)$

$$\dot{\mathbf{x}}(t) = \left[ \mathbf{A}(t) - \mathbf{B}(t) \mathbf{R}^{-1}(t) \mathbf{B}^T(t) \mathbf{P}(t) \right] \mathbf{x}(t)$$

# Linear quadratic optimal control

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(2) When  $t_f$  is infinite ( $t_f \rightarrow \infty$ )

Linear Time-Invariant state space model  $\dot{x}(t) = Ax(t) + Bu(t), \quad x(t_0) = x_0$   
 $y(t) = Cx(t)$

The functional performance  $J = \frac{1}{2} \int_{t_0}^{t_f} [x^T(t) Q x(t) + u^T(t) R u(t)] dt$

where  $Q \geq 0, R > 0$ , Try to find  $u^*$   $J \rightarrow \min$

If (A,B) controllable, and (A,D) observable ( $DD^T = Q$ ),

then there exists the optimal control law  $u^*(t) = -R^{-1} B^T P x(t)$

Optimal performance  $J^* = \frac{1}{2} x^T(0) P x(0)$

where  $P$  satisfies Ricatti equation

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

$$u^*(t) = -R^{-1} B^T P x(t)$$

# Linear quadratic optimal control

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**Example: (P567)** Consider a system  $\dot{\mathbf{x}} = \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix} \mathbf{x} + \begin{bmatrix} 1 \\ 0 \end{bmatrix} u,$

$$\mathbf{x}(0) = \begin{bmatrix} x_1(0) \\ x_2(0) \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

Try to find the optimal control  $u^*$  **that can achieve optimal performance**

$$\mathbf{J} = \int_0^\infty \left[ x_2^2 + \frac{1}{4} u^2 \right] dt$$

The final time is infinite  $\mathbf{J} = \int_0^\infty \left[ x_2^2 + \frac{1}{4} u^2 \right] dt = \frac{1}{2} \int_0^\infty \left[ [x_1, x_2] \begin{bmatrix} 0 & 0 \\ 0 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \frac{1}{2} u^2 \right] dt$

$$Q = \begin{bmatrix} 0 & 0 \\ 0 & 2 \end{bmatrix}, \quad r = \frac{1}{2} \quad DD^T = Q = \begin{bmatrix} 0 & 0 \\ 0 & 2 \end{bmatrix}, \quad D = [0 \quad \sqrt{2}]$$

**Check (A,B) controllable, and (A,D) observable.**

then there exists the optimal control law to make the closed-loop system stable.



# Pole placement for LTI system

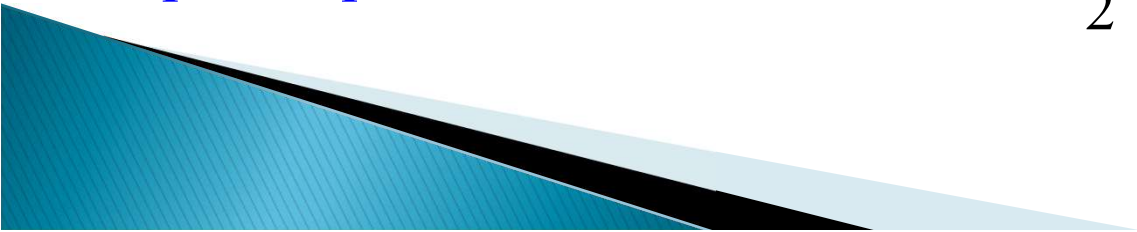
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$$\mathbf{P} = \begin{bmatrix} P_{11} & P_{12} \\ P_{12} & P_{22} \end{bmatrix} \quad \text{Get } \mathbf{P} \text{ from Ricatti equation}$$
$$\mathbf{P}\mathbf{A} + \mathbf{A}^T \mathbf{P} - \mathbf{P}\mathbf{B}r^{-1}\mathbf{B}^T \mathbf{P} + \mathbf{Q} = 0$$

$$\begin{aligned} P_{12} - P_{11}^2 &= 0 \\ P_{22} - 2P_{11}P_{12} &= 0 \\ -2P_{12}^2 + 2 &= 0 \end{aligned} \quad \Rightarrow \mathbf{P} = \begin{bmatrix} 1 & 1 \\ 1 & 2 \end{bmatrix} > 0$$

$$\mathbf{u}^*(t) = -r^{-1}\mathbf{B}^T \mathbf{P}\mathbf{x}(t) = -2x_1(t) - 2x_2(t)$$

Optimal performance

$$J^* = \frac{1}{2} \mathbf{x}^T(0) \mathbf{P}\mathbf{x}(0) = 1$$


# Linear quadratic optimal control

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In MATLAB, there are some functions that can solve linear quadratic problem: `lqr()`、`lqr2()`、`lqry()`

$$J = \frac{1}{2} \int_{t_0}^{t_f} \left[ \mathbf{x}^T(t) \mathbf{Q} \mathbf{x}(t) + \mathbf{u}^T(t) \mathbf{R} \mathbf{u}(t) + 2 \mathbf{x}^T(t) \mathbf{N} \mathbf{u}(t) \right] dt$$

$$[K, S, E] = \text{lqr}(A, B, Q, R, N)$$

**K** is the feedback gain, that  $\mathbf{u} = -\mathbf{K}\mathbf{x}$  ;

**S** the solution P of the associated algebraic Riccati equation;

**E** is the closed-loop eigenvalues,  $E = \text{EIG}(A - B^*K)$ .

$$J = \frac{1}{2} \int_{t_0}^{t_f} \left[ \mathbf{y}^T(t) \mathbf{Q} \mathbf{y}(t) + \mathbf{u}^T(t) \mathbf{R} \mathbf{u}(t) + 2 \mathbf{y}^T(t) \mathbf{N} \mathbf{u}(t) \right] dt$$

$$[K, S, E] = \text{lqry}(A, B, Q, R, N)$$

linear-quadratic (LQ) state-feedback  
regulator with output weighting

# Linear quadratic optimal control

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**Example: Try to find the optimal control law by LQR and LQRY**

$$\dot{x} = \begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ -1 & -4 & -6 \end{pmatrix} x + \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} u$$
$$y = (1 \quad 0 \quad 0) x$$

**(1) lqr()**

$$J = \frac{1}{2} \int_0^\infty (x^T Q x + u^T R u) dt \quad Q = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}, R = 1$$

**(2) lqry()**

$$J = \frac{1}{2} \int_0^\infty (y^T Q y + u^T R u) dt, Q = R = 1$$


# Linear quadratic optimal control

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(1) `lqr()`  $J = \frac{1}{2} \int_0^\infty (x^T Q x + u^T R u) dt$   $Q = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}, R = 1$

$$A = [0, 1, 0; 0, 0, 1; -1, -4, -6];$$

$$B = [0, 0, 1]'; C = [1, 0, 0]; D = 0;$$

$$Q = \text{diag}([1, 1, 1]);$$

$$R = 1;$$

$$K = \text{lqr}(A, B, Q, R)$$

$$A_c = A - B * K; B_c = B; C_c = C; D_c = D;$$

$$\text{Step}(A_c, B_c, C_c, D_c)$$

ans:

$$K = 0.4142 \quad 0.7486 \quad 0.2046$$





# Linear quadratic optimal control

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(2) lqry0

$$J = \frac{1}{2} \int_0^{\infty} (y^T Q y + u^T R u) dt, Q = R = 1$$
$$A = [0, 1, 0; 0, 0, 1; -1, -4, -6];$$
$$B = [0, 0, 1]'; C = [1, 0, 0]; D = 0;$$
$$Q = 1;$$
$$R = 1;$$
$$K = \text{lqry}(A, B, C, D, Q, R)$$
$$A_c = A - B * K; B_c = B; C_c = C; D_c = D;$$
$$\text{Step}(A_c, B_c, C_c, D_c)$$

