

Discrete-time Linear Quadratic Optimal Control

Big picture

Example

Convergence of finite-time LQ solutions

Big picture

- ▶ previously: dynamic programming and **finite-horizon** discrete-time LQ
- ▶ this lecture: infinite-horizon discrete-time LQ and its properties

Review: solution of the general discrete-time LQ problem

- ▶ system dynamics:

$$x(k+1) = A(k)x(k) + B(k)u(k), \quad x(k_0) = x_o \quad (1)$$

- ▶ performance index:

$$J = \frac{1}{2}x^T(N)Sx(N) + \frac{1}{2}\sum_{k=k_0}^{N-1} \left\{ x^T(k)Q(k)x(k) + u^T(k)R(k)u(k) \right\}$$

$$Q(k) = C^T(k)C(k) \succeq 0, \quad S = S^T \succeq 0, \quad R(k) = R^T(k) \succ 0$$

- ▶ optimal $J^o = \frac{1}{2}x_o^T P(0)x_o$ achieved by the state-feedback control law:

$$u^o(k) = - \left[R(k) + B^T(k)P(k+1)B(k) \right]^{-1} B^T(k)P(k+1)A(k)x(k)$$

- ▶ Riccati equation:

$$P(k) = A^T(k)P(k+1)A(k) + Q(k)$$

$$- A^T(k)P(k+1)B(k) \left[R(k) + B^T(k)P(k+1)B(k) \right]^{-1} B^T(k)P(k+1)A(k)$$

with the boundary condition $P(N) = S$.

Example: double integrator

- ▶ plant dynamics:

$$\begin{bmatrix} x_1(k+1) \\ x_2(k+1) \end{bmatrix} = \begin{bmatrix} 1 & T \\ 0 & 1 \end{bmatrix} \begin{bmatrix} x_1(k) \\ x_2(k) \end{bmatrix} + \begin{bmatrix} \frac{T^2}{2} \\ T \end{bmatrix} u(k), \quad T = 1$$
$$y(k) = \begin{bmatrix} 1 & 0 \end{bmatrix} \begin{bmatrix} x_1(k) \\ x_2(k) \end{bmatrix}$$

- ▶ performance index:

$$J_N(x_0) = \frac{1}{2} x^T(N) S x(N) + \frac{1}{2} \sum_{k=0}^{N-1} \left\{ x^T(k) Q x(k) + R u^2(k) \right\}$$

where

$$Q = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}, \quad R > 0$$

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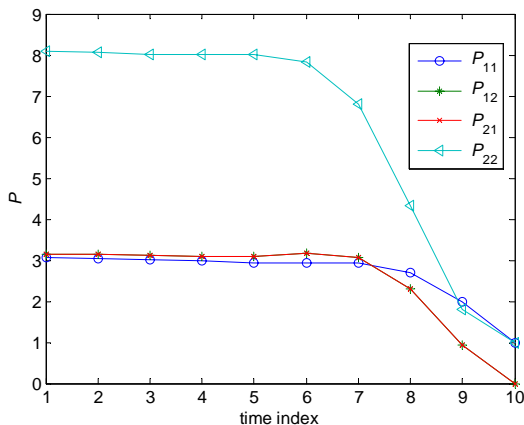
$$Q = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}, \quad R > 0$$

- ▶ next: examine the convergence of $P(k)$ with different $P(N) = S (\succeq 0)$

$$P(k) = A^T P(k+1) A + Q - A^T P(k+1) B \left[R + B^T P(k+1) B \right]^{-1} B^T P(k+1) A$$

Example: double integrator

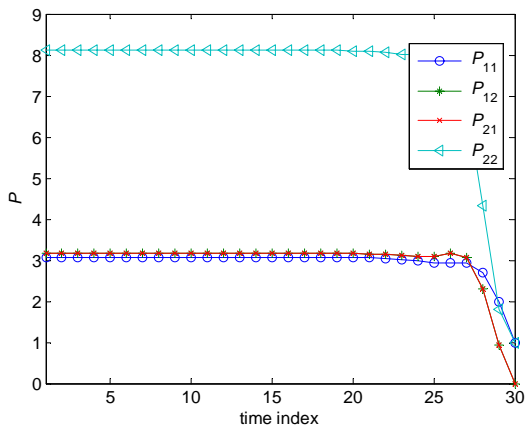
► $N = 10$, $P(N) = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$



$$P(k) = A^T P(k+1) A + Q - A^T P(k+1) B \left[R + B^T P(k+1) B \right]^{-1} B^T P(k+1) A$$

Example: double integrator

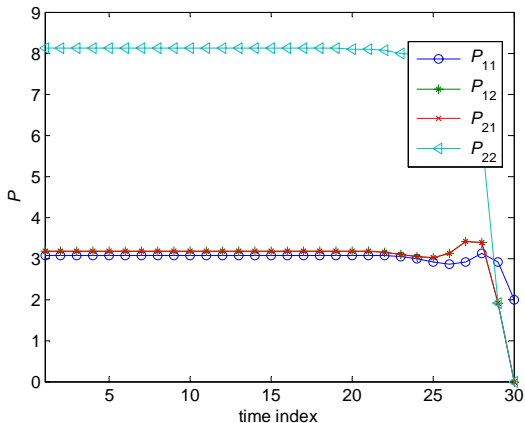
► $N = 30$, $P(N) = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$



$$P(k) = A^T P(k+1) A + Q - A^T P(k+1) B \left[R + B^T P(k+1) B \right]^{-1} B^T P(k+1) A$$

Example: double integrator

► $N = 30$, $P(N) = \begin{bmatrix} 2 & 0 \\ 0 & 0 \end{bmatrix}$



$$P(k) = A^T P(k+1) A + Q - A^T P(k+1) B \left[R + B^T P(k+1) B \right]^{-1} B^T P(k+1) A$$

Example: double integrator

observations:

- ▶ $P(k)$ is indeed always symmetric
- ▶ regardless of the boundary condition $P(N) (\succeq 0)$, the solution of the Riccati equation converges to the same steady state P_s

Example: double integrator

observations:

- ▶ $P(k)$ is indeed always symmetric
- ▶ regardless of the boundary condition $P(N)(\succeq 0)$, the solution of the Riccati equation converges to the same steady state P_s
- ▶ the control law

$$u^o(k) = - \left[R + B^T P(k+1) B \right]^{-1} B^T P(k+1) A x(k)$$

thus converges (backwards) to

$$u^o(k) = - \underbrace{\left(R + B^T P_s B \right)^{-1} B^T P_s A}_{K_s} x(k)$$

From finite-horizon to infinite-horizon LQ

in the case of $N \rightarrow \infty$, it turns out that

- ▶ (A, B) is controllable or stabilizable \Rightarrow guaranteed convergence of $P(k)$ to a bounded P_s
- ▶ intuition: if (A, B) is unstabilizable, then there are unstable uncontrollable modes that may cause

$$\lim_{N \rightarrow \infty} J_N(x_0) = \lim_{N \rightarrow \infty} \left\{ \frac{1}{2} x^T(N) S x(N) + \frac{1}{2} \sum_{k=0}^{N-1} \left\{ x^T(k) Q x(k) + u^T(k) R(k) u(k) \right\} \right\} = \infty$$

$$\text{yielding } J_N^o(x_0) = \frac{1}{2} x_o^T P(0) x_0 = \infty \Rightarrow \lim_{N \rightarrow \infty} \|P(0)\| = \infty$$

Infinite-horizon discrete-time LQ for LTI systems

- ▶ system dynamics:

$$x(k+1) = Ax(k) + Bu(k), \quad x(k_0) = x_0 \quad (2)$$

- ▶ performance index:

$$J = \frac{1}{2} \sum_{k=k_0}^{\infty} \left\{ x^T(k) Q x(k) + u^T(k) R u(k) \right\}, \quad Q \succeq 0, \quad R \succ 0$$

- ▶ optimal state-feedback control law:

$$u^o(k) = - \underbrace{\left(R + B^T P_s B \right)^{-1} B^T P_s A}_{K_s} x(k)$$

- ▶ Algebraic Riccati equation:

$$P_s = A^T P_s A + Q - A^T P_s B \left(R + B^T P_s B \right)^{-1} B^T P_s A$$

Infinite-horizon discrete-time LQ for LTI systems

conditions for a meaningful solution:

$$(A, B) \text{ controllable/stabilizable and } (A, C) \text{ observable/detectable} \\ (Q = C^T C) \Rightarrow \text{guaranteed closed-loop asymptotic stability for} \\ x(k+1) = (A - BK_s)x(k) \triangleq A_{cl}x(k)$$

“(A, B) controllable/stabilizable $\Rightarrow P_s$ and K_s bounded”: already shown
“observability $\Rightarrow P_s \succ 0$ ”: with $u^o(k) = -K_s x(k)$ and $Q = C^T C$

$$\begin{aligned} x_o^T P_s x_o &= \sum_{k=0}^{\infty} \left\{ x^T(k) Q x(k) + u^T(k) R u(k) \right\} = \sum_{k=0}^{\infty} \left\{ x^T(k) \begin{bmatrix} C \\ R^{1/2} K_s \end{bmatrix}^T \begin{bmatrix} C \\ R^{1/2} K_s \end{bmatrix} x(k) \right\} \\ &= \sum_{k=0}^{\infty} \left\{ x_o^T (A_{cl}^k)^T \begin{bmatrix} C \\ R^{1/2} K_s \end{bmatrix}^T \begin{bmatrix} C \\ R^{1/2} K_s \end{bmatrix} A_{cl}^k x_o \right\} = x_o^T W_{cl} x_o \end{aligned}$$

where W_{cl} is the observability gramian for

$$x(k+1) = (A - BK_s)x(k) = A_{cl}x(k) \quad (3)$$

$$\tilde{y}(k) = \begin{bmatrix} C \\ R^{1/2} K_s \end{bmatrix} x(k)$$

Infinite-horizon discrete-time LQ for LTI systems

“observability $\Rightarrow P_s \succ 0$ ” (continued):

- ▶ observability is invariant under static output feedback control \Rightarrow

$$x(k+1) = (A - BK_s)x(k) = A_{cl}x(k)$$

$$\tilde{y}(k) = \begin{bmatrix} C \\ R^{1/2}K_s \end{bmatrix} x(k)$$

is observable if the open-loop system

$$x(k+1) = Ax(k) + Bu(k)$$

$$\tilde{y}(k) = \begin{bmatrix} C \\ R^{1/2}K_s \end{bmatrix} x(k)$$

is observable (which holds as (A, C) is observable). Hence $P_s = W_{cl}$ is positive definite under observability. Analogous analysis can be applied to the detectability case.

Infinite-horizon discrete-time LQ for LTI systems

closed-loop stability of

$$\begin{cases} x(k+1) &= (A - BK_s)x(k) = A_{cl}x(k) \\ \tilde{y}(k) &= \begin{bmatrix} C \\ R^{1/2}K_s \end{bmatrix} x(k) \end{cases} \quad (4)$$

comes from a transformation from Riccati equation to Lyapunov equation:

$$\begin{aligned} P_s &= A^T P_s A + Q - \underbrace{A^T P_s B (R + B^T P_s B)^{-1} B^T P_s A}_{K_s^T} \\ &= \underbrace{A^T P_s A + Q}_{K_s^T} - \underbrace{A^T P_s B (R + B^T P_s B)^{-1} (R + B^T P_s B)}_{K_s} \underbrace{(R + B^T P_s B)^{-1} B^T P_s A}_{K_s} \\ &= \underbrace{(A - BK_s)^T P_s (A - BK_s) + 2A^T P_s BK_s - K_s^T B^T P_s BK_s + C^T C - K_s^T (R + B^T P_s B) K_s}_{K_s^T} \\ &= (A - BK_s)^T P_s (A - BK_s) + C^T C + K_s^T R K_s \\ &\iff \boxed{(A - BK_s)^T P_s (A - BK_s) - P_s = - \begin{bmatrix} C \\ R^{1/2} K_s \end{bmatrix}^T \begin{bmatrix} C \\ R^{1/2} K_s \end{bmatrix} \succeq 0} \quad (5) \end{aligned}$$

observability of (4) plus $P_s \succ 0 \Rightarrow$ closed-loop stability from (5)

Remark

Theorem (An extension of Lyapunov theory based on observability).

if we find from a Lyapunov equation $A^T P A - P = -Q$ where $P \succ 0$, $Q = C^T C \succeq 0$, and (A, C) is observable, then the system $x(k+1) = Ax(k)$ is asymptotically stable.

Proof: Since $P \succ 0$ and $Q \succeq 0$, the system is stable in the sense of Lyapunov. All eigenvalues of A are hence on or inside the unit circle. Pick any eigenvalue-eigenvector pair (λ, v) where λ is on the unit circle. Then $\|Cv\|_2^2 = v^* Q v = -v^* (A^T P A - P) v = -(|\lambda|^2 - 1) v^* P v = 0$, which implies $Cv = 0$. We thus have

$$\begin{cases} Av &= \lambda v \\ Cv &= 0 \\ CAv &= \lambda Cv = 0 \\ \vdots & \\ CA^{n-1}v &= 0 \end{cases} \Rightarrow \begin{bmatrix} C \\ CA \\ \vdots \\ CA^{n-1} \end{bmatrix} v = 0 \Leftrightarrow \begin{bmatrix} C \\ CA \\ \vdots \\ CA^{n-1} \end{bmatrix} \text{ has to be full-column rank}$$

Infinite-horizon discrete-time LQ for LTI systems

an example of closed-loop stability requirement: consider

$$x(k+1) = 2x(k) + u(k), \quad Q = 0, \quad R = 1$$

- ▶ the state constraint Q is zero, the optimal control is $u^o(k) = 0$
- ▶ the closed-loop system is thus unstable
- ▶ on the other hand, (A, C) is clearly unobservable
- ▶ the Riccati equation however still converges, as (A, B) is controllable

Infinite-horizon discrete-time LQ for LTI systems

Excellent closed-loop properties

- ▶ guaranteed closed-loop stability (just shown moments ago)
- ▶ good margins for single-input systems (see book):

$$\text{Phase margin} > 2 \sin^{-1} \left(\frac{1}{2} \sqrt{\frac{R}{R + B^T P_s B}} \right)$$

$$\text{Gain margin} > \frac{1}{1 - \sqrt{\frac{R}{R + B^T P_s B}}}$$

- ▶ guaranteed stability for “% loop gain change” (see book):

$$\frac{100}{1 + \sqrt{\frac{R}{R + B^T P_s B}}} < \text{loop gain change} < \frac{100}{1 - \sqrt{\frac{R}{R + B^T P_s B}}}$$

Summary

- 1 Review: general discrete-time LQ problem
- 2 Example: double integrator
- 3 Convergence of the Riccati equation solution
- 4 Infinite-horizon discrete-time LQ and its properties