

Optimal Control

Lecture 5

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Reading: Sections 2.1 and 2.2 from Ref.[2]

Optimal control of multi-stage systems over finite horizon

Minimum energy control for linear LTI systems with fix final state

$$\begin{aligned} \mathbf{u}^* = \operatorname{argmin} & \frac{1}{2} \sum_{k=0}^{N-1} \mathbf{u}_k^T \mathbf{R}_k \mathbf{u}_k, \text{ s.t.} \\ & \mathbf{x}_{k+1} = \mathbf{A} \mathbf{x}_k + \mathbf{B} \mathbf{u}_k, \\ & \mathbf{x}_0 = \mathbf{x}_0, \quad \mathbf{x}_N = \mathbf{r}_N. \end{aligned}$$

$$\mathbf{u}_k^* = \mathbf{R}^{-1} \mathbf{B}^T (\mathbf{A}^T)^{N-k-1} \mathbf{G}_{0,N}^{-1} (\mathbf{r}_N - \mathbf{A}^N \mathbf{x}_0).$$

where

$$\begin{aligned} \mathbf{G}_{0,N} &= \sum_{i=0}^{N-1} \mathbf{A}^{N-i-1} \mathbf{B} \mathbf{R}^{-1} \mathbf{B}^T (\mathbf{A}^T)^{N-i-1} \\ &= \underbrace{[\mathbf{B} \quad \mathbf{A}\mathbf{B} \quad \dots \quad \mathbf{A}^{N-1}\mathbf{B}]}_{\mathbf{U}_N} \begin{bmatrix} \mathbf{R}^{-1} & & 0 \\ & \ddots & \\ 0 & & \mathbf{R}^{-1} \end{bmatrix} [\mathbf{B} \quad \mathbf{A}\mathbf{B} \quad \dots \quad \mathbf{A}^{N-1}\mathbf{B}]^T \end{aligned}$$

- if $|\mathbf{R}| \neq 0$, solution exists ($\mathbf{G}_{0,N}$ is invertible) if system is reachable (\mathbf{U}_N is full rank)
 - $N \geq n$ (recall Cayley-Hamilton theorem)
- This is an **open-loop** control (depends only on \mathbf{r}_N and \mathbf{x}_0)
 - If system deviates, there is no way to notice the deviation and respond to it. This is not a robust controller.

Couple of points about this problem.

* If there was no control $x(k+1) = Ax(k)$, then

$x_k = A^k x_0 \Rightarrow x_N = A^N x_0$. Thus $r_N - A^N x_0$ is the difference between the desired and undriven final state, it makes sense that u_k^* should depend on this quantity.

* This is an open-loop control. It can be pre-computed knowing only the given x_0 and desired r_N and it is independent of the values of x_k in $k \in [0, N]$. This means that if we apply u_k^* as calculated to the actual system all is well as long as the system model is accurate and nothing occurs to cause x_k to deviate from the optimal trajectory. In practice, however nature is seldom cooperative. Modeling uncertainties and noise cause errors in x_k and the control does not take these errors into account.

* To compute $G_{0,N}$ we can use the alternative approach. The solution to the Lyapunov equation

$$P_{k+1} = AP_k + BR^{-1}B^T, \quad k > 0$$

$$\text{is } P_k = A^k P_0 (A^k)^T + \sum_{i=0}^{k-1} A^{k-i-1} B R^{-1} B^T (A)^{k-i-1}$$

So if we solve this equation with $P_0 = 0$, then $G_{0,k} = P_k$ for each k . First this recursion is solved to compute $G_{0,N}$ and then u_k^*

Does this control exist?

$G_{0,N}$ should be invertible!

$G_{0,N}$ is the weighted reachability gramian of the system.

In terms of the system reachability matrix $U_k = [B \ AB \ \dots \ A^{k-1}B]$

it can be written as

$$G_{0,N} = U_N \begin{bmatrix} R^{-1} & & & 0 \\ & \ddots & & \\ & & \ddots & \\ 0 & & & R^{-1} \end{bmatrix} U_N^T$$

If $R=I$, then $G_{0,N} = U_N U_N^T$. The optimal control exists iff $|G_{0,N}| \neq 0$ since we assume $|R| \neq 0$, this is equivalent to U_N having full rank n , where n is the state dimension ($x \in \mathbb{R}^n$). Therefore, we can drive any x_0 to any desired $x_N = r_N$ for some N iff the system is reachable. Since reachability implies U_{N+j} has

full rank for all $j \geq 0$, if the system is reachable,

we can guarantee the existence of a control to drive x_0 to

$x_N = r_N$ for any r_N by selecting $N \geq n$.

Bryson (1975)

Sweeping method

$$\lambda_k = S_k x_k \leftarrow k=0, \dots, N-1$$

Assume

Lets see if this consistent with the optimality conditions.

$$\left\{ \begin{aligned} * R_k u_k + B_k^T \lambda_{k+1} = 0 &\Rightarrow u_k^* = -R_k^{-1} B_k^T \lambda_{k+1}, \quad k=0, \dots, N-1 \\ u_k = -R_k^{-1} B_k^T \lambda_{k+1} &= -R_k^{-1} B_k^T S_{k+1} x_{k+1} \end{aligned} \right.$$

$$\left\{ \begin{aligned} x_{k+1} &= A_k x_k + B_k (-R_k^{-1} B_k^T S_{k+1}) x_{k+1} \\ x_{k+1} &= (I + B_k R_k^{-1} B_k^T S_{k+1})^{-1} A_k x_k \end{aligned} \right.$$

$$\left\{ \begin{aligned} \text{costate eq } \lambda_k &= Q_k x_k + A_k^T \lambda_{k+1} \quad k=1, \dots, N-1 \\ S_k x_k &= Q_k x_k + A_k^T S_{k+1} x_{k+1} \end{aligned} \right.$$

$$S_k x_k = Q_k x_k + A_k^T S_{k+1} (I + B_k R_k^{-1} B_k^T S_{k+1})^{-1} A_k x_k$$

has to hold for any x_k
and usually $x_k \neq 0$

$$S_k = Q_k + A_k^T S_{k+1} (I + B_k R_k^{-1} B_k^T S_{k+1})^{-1} A_k$$

matrix inversion Lemma $(A + UCV)^{-1} = A^{-1} - A^{-1}U(C^{-1} + VA^{-1}U)^{-1}VA^{-1}$

$$S_k = A_k^T (S_{k+1}^{-1} + B_k R_k^{-1} B_k^T)^{-1} A_k + Q_k$$

$\rightarrow \{S_k\}_{k=1}^{N-1}$

$S_N > 0, Q_k > 0 \rightarrow S_k > 0$

↳ this can be computed off-line!

$$\lambda_k = S_k x_k \quad k=1, \dots, N$$

$$U_k^* = -R_k^{-1} B_k^T \lambda_{k+1} = -R_k^{-1} B_k^T S_{k+1} x_{k+1}$$

$$U_k^* = - \underbrace{(B_k^T S_{k+1} B_k + R_k)^{-1} B_k^T S_{k+1} A_k}_{K_k} x_k$$

Kalman Gain $\leftarrow K_k \leftarrow$ Can be computed off-line

$$U_k^* = -K_k x_k$$

closed-loop control

↳ Linear Quadratic Regulator (LQR)

controller

* Robust to deviation from optimal trajectory (modeled one)

LQR (finite time)

As good
as possible
*
J as
small
as possible

$$u_k = -k_k x_k \Rightarrow x_{k+1} = (A_k - B_k k_k) x_k$$

@ N: $S_N \leftarrow$ is known

$$k_k = (B_k^T S_{k+1} B_k + R_k)^{-1} B_k^T S_{k+1} A_k$$

$$S_k = A_k^T S_{k+1} (A_k - B_k k_k) + Q_k$$

Joseph stabilized form

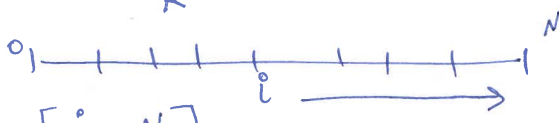
$$S_k = (A - B_k k_k)^T S_{k+1} (A - B_k k_k) + k_k^T R_k k_k + Q_k$$

Even if (A, B) time-invariant, optimal gain

$\{k_k\}_{k=0}^{N-1}$ is time varying.

* Can I use a suboptimal control with fixed gain?

$$J = \frac{1}{2} x_N^T S_N x_N + \frac{1}{2} \sum_{k=0}^{N-1} x_k^T Q_k x_k + u_k^T R_k u_k$$

cost to go 

$$J_i = \frac{1}{2} x_N^T S_N x_N + \frac{1}{2} \sum_{k=i}^{N-1} x_k^T Q_k x_k + u_k^T R_k u_k$$

observe

$$\frac{1}{2} \sum_{k=i}^{N-1} (x_{k+1}^T S_{k+1} x_{k+1} - x_k^T S_k x_k) = \frac{1}{2} x_N^T S_N x_N - \frac{1}{2} x_i^T S_i x_i$$

$$J_i^* = \frac{1}{2} x_i^T S_i x_i$$

once you have S_i , then you have J_i^*