

Problem 1. [100 Points] Platooning

In a single lane straight road, N vehicles are moving to the right with respective 1D position coordinates x_1, x_2, \dots, x_N . See Fig. showing an example scenario for $N = 4$.



Suppose that all vehicles have identical discrete time controlled dynamics $x_i(k+1) = x_i(k) + h(u_i(k) - v)$, $i = 1, 2, \dots, N$, for time index $k = 0, \dots, T-1$. The parameter $h > 0$ is given constant sampling time. The control u_i can be thought of as the speed of the i th vehicle, and v is a known speed limit.

Here is the high level question of interest: what should be the optimal controls such that all consecutive vehicles maintain a separation close to some (known) desired distance d at all times? A separation smaller than d may be unsafe, and thus undesirable. A separation more than d reduces traffic throughput, and therefore also undesirable. We also want all vehicles to move at a speed close to the known speed limit v .

(a) [35 points] OCP formulation

Motivated by the aforesaid objective, consider minimizing

$$\frac{1}{2} \sum_{i=1}^{N-1} (x_{i+1}(T) - x_i(T) - d)^2 + \frac{1}{2} \sum_{k=0}^{T-1} \left\{ \sum_{i=1}^{N-1} (x_{i+1}(k) - x_i(k) - d)^2 + \sum_{i=1}^N (u_i(k) - v)^2 \right\}$$

subject to $x_i(k+1) = x_i(k) + h(u_i(k) - v)$, $i = 1, 2, \dots, N$. Consider the final time T fixed.

Recast this problem as discrete time finite horizon LQ tracking by clearly defining the **state vector** x and its **dimension**, the **control vector** u and its **dimension**, the **output vector** y and its **dimension**, the **desired output trajectory** y_d to track, the **system matrices** A, B, C , and the **weight matrices** M, Q, R in the **cost function**.

Hint: Take a look at Lec. 10, p. 3 to see how LQ tracking problem was formulated in the continuous time case. See also part (b) to get a hint on the problem structure.

Solution for part (a):

Let $\mathbf{1}_n$ denote an $n \times 1$ column vector of ones, and let I_n denote the $n \times n$ identity matrix. We define the state vector x , the control vector u , the output vector y , and the desired output vector y_d as

$$x := \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_N \end{pmatrix} \in \mathbb{R}^N, \quad u := \begin{pmatrix} u_1 - v \\ u_2 - v \\ \vdots \\ u_N - v \end{pmatrix} \in \mathbb{R}^N, \quad y := \begin{pmatrix} x_2 - x_1 \\ x_3 - x_2 \\ \vdots \\ x_N - x_{N-1} \end{pmatrix} \in \mathbb{R}^{N-1}, \quad y_d := d\mathbf{1}_{N-1} \in \mathbb{R}^{N-1}.$$

Consequently, the system matrices are

$$A := I_N, \quad B := hI_N, \quad C := \begin{pmatrix} -1 & 1 & 0 & 0 & \dots & 0 \\ 0 & -1 & 1 & 0 & \dots & 0 \\ 0 & 0 & -1 & 1 & \dots & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & \dots & 0 & -1 & 1 \end{pmatrix} \in \mathbb{R}^{(N-1) \times N}.$$

The cost weight matrices are

$$M = Q = I_{N-1} > 0, \quad R = I_N > 0.$$

With the above definitions in place, we can transcribe the given problem into standard discrete time LQR tracking problem:

$$\underset{\{u_k\}_{k=0}^{T-1}}{\text{minimize}} \quad \frac{1}{2} \left\{ (y(T) - y_d(T))' M (y(T) - y_d(T)) + \sum_{k=0}^{T-1} (y(k) - y_d(k))' Q (y(k) - y_d(k)) + (u(k))' R u(k) \right\}$$

subject to $x(k+1) = Ax(k) + Bu(k)$, $y(k) = Cx(k)$, $k = 0, 1, \dots, T-1$.

(b) [35 points] Discrete time LQ tracking solution

Extend the derivation in Lec. 10, p. 15-23 for the tracking case:

$$\underset{\{u_k\}_{k=0}^{T-1}}{\text{minimize}} \quad \frac{1}{2} \left\{ (y(T) - y_d(T))' M (y(T) - y_d(T)) + \sum_{k=0}^{T-1} (y(k) - y_d(k))' Q (y(k) - y_d(k)) + (u(k))' R u(k) \right\}$$

subject to $x(k+1) = Ax(k) + Bu(k)$, $y(k) = Cx(k)$, $k = 0, 1, \dots, T-1$.

Hint: Just like the continuous time LQ tracking solution given in Lec. 10, here too you should get optimal control as a sum of a linear state feedback term, and a feedforward term. You need to derive a backward **vector** recursion for the feedforward control. Also derive how the Riccati backward recursion needs to be modified in this case, compared to the same for LQR.

Solution for part (b):

As usual, we start with the Hamiltonian

$$H(k) = \frac{1}{2}(Cx(k) - d\mathbf{1})'Q(Cx(k) - d\mathbf{1}) + \frac{1}{2}(u(k))'Ru(k) + (\lambda(k+1))'(Ax(k) + Bu(k)),$$

which gives the following necessary conditions for optimality:

$$\begin{aligned} x(k+1) &= Ax(k) + Bu(k), \quad \lambda(k) = \frac{\partial H(k)}{\partial x(k)} = C'QCx(k) - C'Qd\mathbf{1} + A'\lambda(k+1), \\ 0 &= \frac{\partial H(k)}{\partial u(k)} = Ru(k) + B'\lambda(k+1), \end{aligned}$$

with boundary conditions

$$\lambda(T) = \frac{\partial \phi(x(T), T)}{\partial x(T)} = C'M(Cx(T) - d\mathbf{1}), \quad x(0) \text{ given.}$$

This gives us the 2PBVP

$$\begin{pmatrix} x(k+1) \\ \lambda(k) \end{pmatrix} = \begin{pmatrix} A & -BR^{-1}B' \\ C'QC & A' \end{pmatrix} \begin{pmatrix} x(k) \\ \lambda(k+1) \end{pmatrix} + \begin{pmatrix} 0 \\ -C'Qd\mathbf{1} \end{pmatrix} d\mathbf{1}$$

subject to the above boundary conditions.

We consider the ansatz $\lambda(k) = P(k)x(k) - v(k)$, which upon comparing with our boundary condition, yields

$$P(T) = C'MC, \quad v(T) = C'Md\mathbf{1}.$$

Now we proceed as in Lec. 10, p. 18-19, mutatis mutandis, to obtain

$$\begin{aligned} \lambda(k+1) &= P(k+1)x(k+1) - v(k+1) = P(k+1)(Ax(k) - BR^{-1}B'\lambda(k+1)) - v(k+1) \\ \Rightarrow \lambda(k+1) &= (I + P(k+1)BR^{-1}B')^{-1}P(k+1)Ax(k) - (I + P(k+1)BR^{-1}B')^{-1}v(k+1). \end{aligned}$$

On the other hand, the costate equation gives $\lambda(k) = C'QCx(k) + A'\lambda(k+1) - C'Qd\mathbf{1}$, wherein the LHS equals $P(k)x(k) - v(k)$, and in the RHS, we substitute the expression for $\lambda(k+1)$ derived above, to arrive at

$$\begin{aligned} P(k)x(k) - v(k) &= \left(C'QC + A'(I + P(k+1)BR^{-1}B')^{-1}P(k+1)A \right) x(k) - A'(I + P(k+1)BR^{-1}B')^{-1}v(k+1) \\ &\quad - C'Qd\mathbf{1}. \end{aligned}$$

Since the above must hold for all initial conditions x_0 , and thus for arbitrary $x(k)$, we must have:

$$\begin{aligned} P(k) &= C'QC + A'(I + P(k+1)BR^{-1}B')^{-1}P(k+1)A, \quad P(T) = C'MC, \\ v(k) &= A'(I + P(k+1)BR^{-1}B')^{-1}v(k+1) + C'Qd\mathbf{1}, \quad v(T) = C'Md\mathbf{1}, \end{aligned}$$

which may be put in slightly more symmetric form as in Lec. 10, p. 20.

Once we solve the above matrix-vector backward recursions, we can write the optimal control u^{opt} as follows:

$$\begin{aligned}
u^{\text{opt}}(k) &\stackrel{(\text{PMP})}{=} -R^{-1} B' \lambda(k+1) = -R^{-1} B' P(k+1) (Ax(k) + Bu^{\text{opt}}(k)) + R^{-1} B' v(k+1) \\
\Rightarrow (I + R^{-1} B' P(k+1) B) u^{\text{opt}}(k) &= -R^{-1} B' P(k+1) Ax(k) + R^{-1} B' v(k+1) \\
\Rightarrow u^{\text{opt}}(k) &= -(I + R^{-1} B' P(k+1) B)^{-1} R^{-1} B' P(k+1) Ax(k) + (I + R^{-1} B' P(k+1) B)^{-1} R^{-1} B' v(k+1) \\
&= -(R^{-1} R + B' P(k+1) B)^{-1} R^{-1} B' P(k+1) Ax(k) + (R^{-1} R + B' P(k+1) B)^{-1} R^{-1} B' v(k+1) \\
&= -(R + B' P(k+1) B)^{-1} R R^{-1} B' P(k+1) Ax(k) + (R + B' P(k+1) B)^{-1} R R^{-1} B' v(k+1) \\
&= -(R + B' P(k+1) B)^{-1} B' P(k+1) Ax(k) + (R + B' P(k+1) B)^{-1} B' v(k+1) \\
&= u_{\text{feedback}} + u_{\text{feedforward}},
\end{aligned}$$

where

$$u_{\text{feedback}} := -K(k)x(k), \quad K(k) := K_v P(k+1)A, \quad K_v := (R + B' P(k+1)B)^{-1} B', \quad u_{\text{feedforward}} := K_v v(k+1).$$

(c) [30 points] Optimal control for platooning

Apply your answer in part (b) to the formulation in part (a), to **compute and plot** $y^{\text{opt}}(k)$ superimposed with $y_d(k)$. Also plot $u^{\text{opt}}(k)$. To make these plots, fix $T = 200$, $h = 0.01$, $N = 4$, $d = 245$ ft, and the initial conditions $x_1(0) = 0$ ft, $x_2(0) = 250$ ft, $x_3(0) = 480$ ft, $x_4(0) = 780$ ft.

Please submit your single MATLAB code generating these plots.

Solution for part (c):

Please see the posted MATLAB code UCSC-AM232-S21-HW4.m posted in CANVAS Files section, folder: HW. The plots are shown below.



