

EE5630: Optimal Control - Assignment 2

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March 1, 2018

Question 1

Given the system dynamics

$$\begin{aligned}\dot{x}_1(t) &= x_2(t) \\ \dot{x}_2(t) &= -x_1(t) - x_2(t) + u(t)\end{aligned}$$

and the cost function as

$$J = \int_0^T \frac{1}{2} [q_1 x_1^2(t) + q_2 x_2^2(t) + u^2(t)] dt; \quad q_1, q_2 > 0.$$

Find $U^*(t)$ expressed as a function of $X(t)$, t , and J_X^* for the given system.

Solution

Let

$$\begin{aligned}\mathcal{H} &= g + J_X^{*\top} [a] \\ &= q_1 x_1^2 + q_2 x_2^2 + u^2 + \begin{bmatrix} J_{x_1} & J_{x_2} \end{bmatrix} \begin{bmatrix} x_2 \\ -x_1 + x_2 + u \end{bmatrix} \\ &= \frac{1}{2} (q_1 x_1^2 + q_2 x_2^2 + u^2) + J_{x_1} x_2 + J_{x_2} (-x_1 + x_2 + u)\end{aligned}$$

Now, find $\frac{\partial \mathcal{H}}{\partial u} = 0$

$$\begin{aligned}\frac{\partial \mathcal{H}}{\partial u} &= u^* + J_{x_2} \\ &= 0 \\ \frac{\partial^2 \mathcal{H}}{\partial u^2} &= 1 > 0 \quad \therefore \text{minimum}\end{aligned}$$

Therefore, the optimal control trajectory is $\boxed{u^*(t) = -J_{x_2}^*(x(t), t)}$.

Question 2

A system has the following first order linear dynamics

$$\dot{x}(t) = -10x(t) + u(t)$$

which needs to be controlled while minimizing the following cost function:

$$J = \frac{1}{2}x^2(T) + \int_0^T [\frac{1}{4}x^2(t) + \frac{1}{2}u^2(t)]dt;$$

The admissible state and control trajectories are not constrained by any boundaries. Find the optimal control law using HJB equation for $T = 5$ and assuming the solution of HJB PDE to be of the quadratic form. You can leave your answers in terms for $K(t)$.

Solution

Let

$$\begin{aligned}\mathcal{H} &= g + J_X^{*\top}[a] \\ &= (\frac{1}{4}x^2 + \frac{1}{2}u^2) + J_x(-10x + u)\end{aligned}$$

Now, find $\frac{\partial \mathcal{H}}{\partial u} = 0$

$$\begin{aligned}\frac{\partial \mathcal{H}}{\partial u} &= u^* + J_x \\ &= 0 \\ u^* &= -J_x \\ \frac{\partial^2 \mathcal{H}}{\partial u^2} &= 1 > 0 \therefore \text{minimum}\end{aligned}$$

This system is Linear, Quadratic, and a Regulator (LQR) problem. Therefore, the optimal control trajectory is $\boxed{u^*(t) = -k(t)x(t)}$. The minimum cost, J^* , of an LQR problem is given by $J^* = \frac{1}{2}k(t)x^2(t)$ and the solution to the Hamilton-Jacobi-Bellman of an LQR problem can be found in the following manner:

1.

$$\begin{aligned}J_t^* &= \frac{1}{2}\dot{k}(t)x(t) \\ J_x^* &= k(t)x(t)\end{aligned}$$

2.

$$\begin{aligned}
\mathcal{H}^* &= \frac{1}{4}x^2 + \frac{1}{2}(-J_x^*)^2 + J_x^*(-10x - J_x^*) \\
&= \frac{1}{4}x^2 + \frac{1}{2}J_x^{*2} - 10J_x^*x - J_x^{*2} \\
&= \frac{1}{4}x^2 - \frac{1}{2}J_x^{*2} - 10J_x^*x
\end{aligned}$$

3.

$$\begin{aligned}
0 &= J_t^* + \mathcal{H}^* \\
&= J_t^* + \frac{1}{4}x^2 - \frac{1}{2}J_x^{*2} - 10J_x^*x \\
&= \frac{1}{2}\dot{k}x^2 + \frac{1}{4}x^2 - \frac{1}{2}(kx)^2 - 10(kx)x \\
&= \frac{1}{2}\dot{k}x^2 + \frac{1}{4}x^2 - \frac{1}{2}k^2x^2 - 10kx^2 \\
&= \dot{k} + \frac{1}{2} - k^2 - 20k
\end{aligned} \tag{1}$$

4. Using Wolfram Alpha and taking the natural log (ln) of Equation 1 yields

$$t + \alpha = \frac{1}{\sqrt{402}} \ln \left[\frac{2k(t) + 20 - \sqrt{402}}{2k(t) + 20 + \sqrt{402}} \right], \quad \alpha \in \mathbb{R}$$

Question 3

The dynamics of a nonlinear scalar system is:

$$\dot{x}(t) = x(t)u(t); \quad x(0) = 1;$$

and the cost function to be minimized is

$$J = x^2(1) + \int_0^1 [x(t)u(t)]^2 dt;$$

Find the optimal feedback solution by solving the corresponding HJB equation. [Hint: First, find the HJB partial differential equation in terms of J_X , J_t . Then using boundary conditions show that the PDE admits a solution that is quadratic in x . Finally integrate the ODE in $K(t)$ to get the feedback solution.]

Solution

Let

$$\begin{aligned}
\mathcal{H} &= g + J_X^{*\top}[a] \\
&= x^2u^2 + J_x x u
\end{aligned}$$

Now, find $\frac{\partial \mathcal{H}}{\partial u} = 0$

$$\begin{aligned}\frac{\partial \mathcal{H}}{\partial u} &= 2x^2 u^* + J_x x \\ &= 0 \\ u^* &= -\frac{J_x}{2x^2} \\ \frac{\partial^2 \mathcal{H}}{\partial u^2} &= 2x^2 > 0 \quad \therefore \text{minimum}\end{aligned}$$

This system is not LQR problem. Therefore, the optimal control trajectory cannot automatically be assumed to be quadratic in x . However, by looking at the initial conditions we can see that the final cost *is* quadratic in x , $J(x(1), 1) = x^2(1)$. This means that it is safe to assume that throughout the process the cost is quadratic in x .

1.

$$\begin{aligned}J_t^* &= \frac{1}{2} \dot{k}(t) x(t) \\ J_x^* &= k(t) x(t)\end{aligned}$$

2.

$$\begin{aligned}\mathcal{H}^* &= x^2 \left(-\frac{J_x}{2x^2} \right)^2 + J_x^* \left(-\frac{J_x}{2x^2} \right) \\ &= J_x^{*2} - \frac{1}{2} J_x^{*2} \\ &= \frac{1}{2} J_x^{*2}\end{aligned}$$

3.

$$\begin{aligned}0 &= J_t^* + \mathcal{H}^* \\ &= J_t^* + \frac{1}{2} J_x^{*2} \\ &= \frac{1}{2} \dot{k} x^2 + \frac{1}{2} (k x)^2 \\ &= \frac{1}{2} \dot{k} x^2 + \frac{1}{2} k^2 x^2 \\ &= \dot{k} + k^2\end{aligned} \tag{2}$$

4. Solving Equation 2 yields

$$\boxed{k(t) = \frac{1}{t + \alpha}, \quad \alpha \in \mathbb{R}}$$

Question 4

Given the system dynamics of a plant:

$$\begin{aligned}x_1(t) &= x_2(t) \\ x_2(t) &= -x_1(t) - 2x_2(t) + u(t)\end{aligned}$$

and the cost function to minimize is

$$J = 10x_1^2(T) + \int_0^T \frac{1}{2}[x_1^2(t) + 2x_2^2(t) + u^2(t)]dt$$

Final time $T = 10$ and the admissible state and control trajectories are not constrained by any boundaries. Find the optimal control law by solving the Riccati Equation numerically.

- Identify A , B , Q , R , H
- Find and plot $K(t)$ using Riccati Equation
- Find and plot optimal control law
- Plot the trajectories of states x_1 and x_2 using control in part c) (Hint: Identify A , B , Q , R , H etc and then use *ODE45* command in matlab to solve the ODE in $K(t)$)

Solution

a)

$$A = \begin{bmatrix} 0 & 1 \\ -1 & -2 \end{bmatrix}, \quad B = \begin{bmatrix} 0 \\ 1 \end{bmatrix}, \quad Q = \begin{bmatrix} 1 & 0 \\ 0 & 2 \end{bmatrix}, \quad R = 1$$

b)

To use MATLAB's `ode()` function, a function for the ODE that we wish to solve must first be written, as shown in Listing 2. As Listing 1 shows, the `ode()` function requires this ODE function as an input.

Listing 1: Script to solve the Riccati equation using the ODE45 solver

```
%% Solve the Riccati Equation numerically
% adapted from:
% http://www.eng.auburn.edu/~tplacek/courses/3600/ode45waterloo.pdf
% https://www.mathworks.com/matlabcentral/answers/94722-how-can-i-solve-
% the-matrix-riccati-differential-equation-within-matlab

%%% Part b) find and plot K(t) using Riccati Equation
A = [ 0, 1;
      -1, -2];
B = [0;
      1];
```

```

Q = [1, 0;
     0, 2];
R = 1;

X0 = [0, 0, 0, 0]';
period = [0, 10];

% ode45 requires that a function be passed as an input
[t1, X] = ode45(@(t, K)riccati(t, K, A, B, Q, R), period, X0);

plot(t1, X)
legend('k1', 'k2', 'k3', 'k4')

```

The API for MATLAB's `ode()` function requires that the ODE function that is being solved accepts a column vector as a parameter and returns a column vector as well. This is why `reshape()` is used at the beginning of the `riccati()` function.

Listing 2: Function to calculate the Riccati equation at one time step

```

function dkdt = riccati(t, K, A, B, Q, R)
% This is the Riccati equation specific to question 4 from assignment 2
% from Linear Systems Analysis (EE5600) at CSULA.
%
% Adapted from:
% https://www.mathworks.com/matlabcentral/answers/94722-how-can-i-solve-the-matrix-riccati-differential-equation-within-matlab
K = reshape(K, size(A));
dkdt = -Q + K*B*(R^-1)*B' - K*A - A'*K;
dkdt = dkdt(:);
end

```

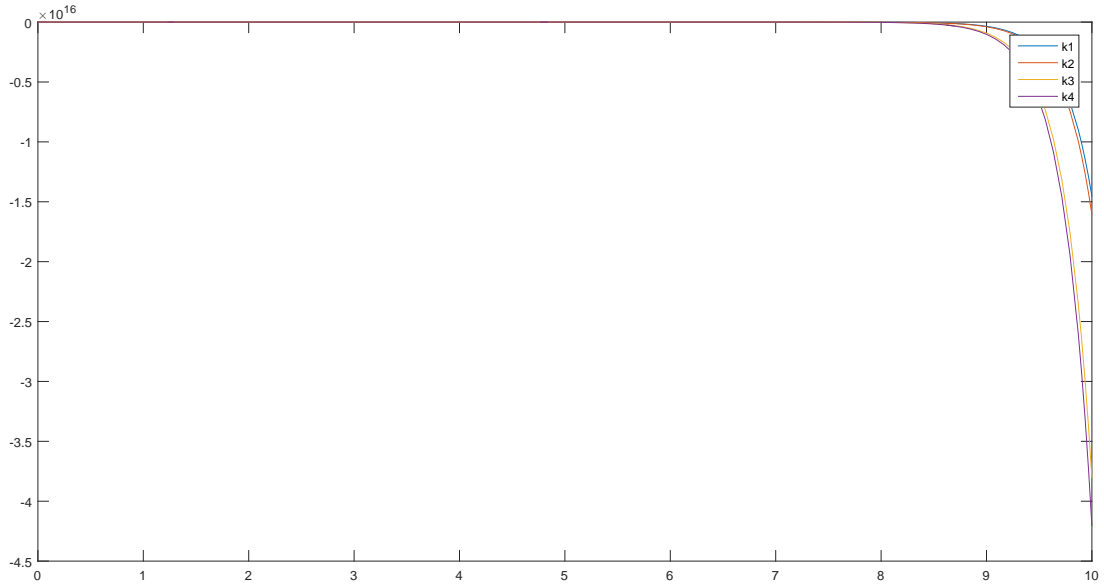


Figure 1: Plot of $K(t)$

c) and d)

The remaining portions of this problem were not completed due to time constraints.

Question 5

Derive the Hamilton-Jacobi-Bellman partial differential equation.

Solution

1. First, the problem must be formulated

(a) System dynamics: $\dot{X}(t) = a(X(t), U(t), t)$

(b) Cost function: $J = h(X(t_f), t_f) + \int_{t_0}^{t_f} g(X(t), U(t), t) dt$

(c) Define the state and control constraints

2. Now, consider a more general cost function (i.e, a *value function*)

$$(a) \quad J(X(t), t, U(\tau)) = h(X(t_f), t_f) + \int_t^{t_f} g(X(\tau), U(\tau), \tau) d\tau$$

for $t \leq \tau \leq t_f$

3. Find the minimum cost function for all admissible $X(t)$ and for all $t \leq \tau \leq t_f$.

$$(a) \quad J^*(X(t), t) = \min_{\substack{U(\tau) \\ t \leq \tau \leq t_f}} \left\{ \int_t^{t_f} g(X(\tau), U(\tau), \tau) d\tau + h(X(t_f), t_f) \right\}$$

4. Subdivide the interval

$$(a) \quad J^*(X(t), t) = \min_{\substack{U(\tau) \\ t \leq \tau \leq t_f}} \left\{ \int_t^{t+\Delta t} g d\tau + \int_{t+\Delta t}^{t_f} g d\tau + h \right\}$$

5. Use the principle of optimality

$$(a) \quad J^*(X(t), t) = \min_{\substack{U(\tau) \\ t \leq \tau \leq t_f}} \left\{ \int_t^{t+\Delta t} g d\tau + J^*(X(t+\Delta t), t+\Delta t) \right\}$$

where $J^*(X(t+\Delta t), t+\Delta t)$ is the minimum cost of the process for the time interval $t+\Delta t \leq \tau \leq t_f$ with initial state $X(t+\Delta t)$.

6. Expand $J^*(X(t+\Delta t), t+\Delta t)$ in a Taylor series about the point $(x(t), t)$ (assuming that the second partial derivatives of J^* exist and are bounded)

(a)

$$\begin{aligned} J^*(X(t), t) = \min_{\substack{U(\tau) \\ t \leq \tau \leq t_f}} \left\{ \int_t^{t+\Delta t} g d\tau + J^*(X(t), t) + \left[\frac{\partial J^*}{\partial t}(X(t), t) \right] \Delta t \right. \\ \left. + \left[\frac{\partial J^*}{\partial X}(X(t), t) \right]^T [X(t+\Delta t) - X(t)] + \mathcal{O}(\Delta t) \right\} \end{aligned}$$

where $\mathcal{O}(\Delta t)$ is higher order terms.

7. For small Δt

(a)

$$\begin{aligned} J^*(X(t), t) = \min_{U(t)} \{ & g(X(t), U(t), t) \Delta t + J^*(X(t), t) \\ & + J_t^*(X(t), t) \Delta t + J_X^{*\top}(X(t), t) [a(X(t), U(t), t)] \Delta t \\ & + \mathcal{O}(\Delta t) \} \end{aligned}$$

8. Remove $J^*(X(t), t)$ and $J_t^*(X(t), t)$ terms (they don't depend on $U(t)$)

(a)

$$0 = J_t^*(X(t), t)\Delta t + \min_{U(t)}\{g(X(t), U(t), t)\Delta t + J_X^{*\top}(X(t), t)[a(X(t), U(t), t)]\Delta t + \mathcal{O}(\Delta t)\}$$

9. Divide by Δt and let $\Delta t \rightarrow 0$

(a)

$$0 = J_t^*(X(t), t) + \min_{U(t)}\{g(X(t), U(t), t) + J_X^{*\top}(X(t), t)[a(X(t), U(t), t)]\}$$

Which can be shortened to

$$0 = J_t^* + \min_{U(t)}\{g + J_X^{*\top}[a]\}$$

or

$$\boxed{0 = J_t^* + \mathcal{H}^*}$$

Question 6

Write a short note on LQR problem and derive the Riccati equation for LQR problem.

Solution

An LQR problem, such as Questions 1 and 2, are problems such that the plant dynamics are linear and the cost function is quadratic. The plant dynamics can be described as in Equation 3 and the cost function to be minimized given in Equation 4.

$$\dot{X}(t) = A(t)X(t) + B(t)U(t) \tag{3}$$

$$J = \frac{1}{2}X^\top(t_f)HX(t_f) + \int_{t_0}^{t_f} \frac{1}{2}[X^\top(t)Q(t)X(t) + U^\top(t)R(t)U(t)]dt \tag{4}$$

where H and Q are real, symmetric positive semi-definite matrices, R is a real, symmetric positive definite matrix, and t_0 and t_f are the initial and final times, respectively, and are specified. The control and state trajectories $U(t)$ and $X(t)$ are not constrained by any boundaries.

1. Form the Hamiltonian

(a)

$$\begin{aligned} \mathcal{H}(X(t), U(t), J_X^*, t) = & \frac{1}{2}[X^\top(t)Q(t)X(t) + U^\top(t)R(t)U(t)] \\ & + J_t^*(X(t), t) \cdot [A(t)X(t) + B(t)U(t)] \end{aligned} \tag{5}$$

2. Minimize the Hamiltonian $\left(\frac{\partial \mathcal{H}}{\partial u} = 0\right)$

(a)

$$\begin{aligned} \frac{\partial \mathcal{H}}{\partial u}(X(t), U(t), J_X^*, t) &= R(t)U(t) + B^\top(t)J_X^*(X(t), t) \\ &= 0 \end{aligned} \quad (6)$$

3. Check that Equation 6 really is a minimum $\left(\frac{\partial^2 \mathcal{H}}{\partial u^2} > 0\right)$

(a)

$$\frac{\partial^2 \mathcal{H}}{\partial u^2} = R(t) > 0$$

because R is real, positive definite

4. Because \mathcal{H} is a quadratic form in $U(t)$, the control trajectory that minimizes Equation 6 is a global minimum. By solving Equation 6 for $U^*(t)$ we get

(a)

$$U^*(t) = -R^{-1}(t)B^\top(t)J_X^*(X(t), t) \quad (7)$$

5. Substitute Equation 7 into Equation 5 (the arguments of the matrices will be dropped from here on due to space constraints)

(a)

$$\begin{aligned} \mathcal{H}(X(t), U(t), J_X^*, t) &= \frac{1}{2}X^\top QX + \frac{1}{2}J_X^{*\top}BR^{-1}B^\top J_X^* \\ &\quad + J_t^*AX - J_X^{*\top}BR^{-1}B^\top J_X^* \\ &= \frac{1}{2}X^\top QX - \frac{1}{2}J_X^{*\top}BR^{-1}B^\top J_X^* - J_t^{*\top}AX \end{aligned}$$

6. The Hamilton-Jacobi-Bellman equation becomes

(a)

$$0 = J_t^* + \frac{1}{2}X^\top QX - \frac{1}{2}J_X^{*\top}BR^{-1}B^\top J_X^* - J_t^{*\top}AX \quad (8)$$

7. The boundary condition

(a)

$$J_X^*(X(t_f), t_f) = \frac{1}{2}X^\top(t_f)HX(t_f)$$

8. Guess a quadratic solution

(a)

$$J_X^*(X(t), t) = \frac{1}{2} X^\top K X$$

where K is a real symmetric positive-definite matrix

9. Substitute our guess into the HJB equation, Equation 8

(a)

$$0 = \frac{1}{2} X^\top \dot{K} X + \frac{1}{2} X^\top Q X - \frac{1}{2} X^\top K B R^{-1} B^\top K X + X^\top K A X \quad (9)$$

10. Use the substitution of $K A = \frac{1}{2} [K A + (K A)^\top] + \frac{1}{2} [K A - (K A)^\top]$, the property that $(C D)^\top = D^\top C^\top$, and the fact that the transpose of a scalar is the scalar to show that only the symmetric part of $K A$ contributes to Equation 9 to get

(a)

$$0 = \frac{1}{2} X^\top \dot{K} X + \frac{1}{2} X^\top Q X - \frac{1}{2} X^\top K B R^{-1} B^\top K X + \frac{1}{2} X^\top K A X + \frac{1}{2} X^\top A^\top K X$$

which holds for X

11. Removing all X

(a) $\boxed{0 = \dot{K} + Q - K B R^{-1} B^\top K + K A + A^\top K}$

12. With the boundary condition

(a) $\boxed{K(t_f) = H}$