

HW3: 2.2-2, 4, 7

2.2-2 Find all possible solutions to (2.2-26):

(2.2-26) $S = A^T S A + Q$

Where $A = \begin{bmatrix} \frac{1}{2} & 1 \\ 0 & -\frac{1}{2} \end{bmatrix}$ $C = \begin{bmatrix} 2 & 0 \end{bmatrix}$ $Q = C^T C$

Let $P = \begin{bmatrix} P_1 & P_2 \\ P_3 & P_4 \end{bmatrix}$, thus (2.2-26) becomes:

$$\begin{bmatrix} P_1 & P_2 \\ P_3 & P_4 \end{bmatrix} = \begin{bmatrix} \frac{1}{2} & 0 \\ 1 & -\frac{1}{2} \end{bmatrix} \begin{bmatrix} P_1 & P_2 \\ P_3 & P_4 \end{bmatrix} \begin{bmatrix} \frac{1}{2} & 1 \\ 0 & -\frac{1}{2} \end{bmatrix} + \begin{bmatrix} 2 \\ 0 \end{bmatrix} \begin{bmatrix} 2 & 0 \end{bmatrix}$$

$$\begin{bmatrix} P_1 & P_2 \\ P_3 & P_4 \end{bmatrix} = \begin{bmatrix} \frac{P_1}{2} & \frac{P_2}{2} \\ \frac{P_1 - P_3}{2} & \frac{P_2 - P_4}{2} \end{bmatrix} \begin{bmatrix} \frac{1}{2} & 1 \\ 0 & -\frac{1}{2} \end{bmatrix} + \begin{bmatrix} 4 & 0 \\ 0 & 0 \end{bmatrix}$$

$$\begin{bmatrix} P_1 & P_2 \\ P_3 & P_4 \end{bmatrix} = \begin{bmatrix} \frac{P_1}{4} + 4 & \frac{P_1}{2} - \frac{P_3}{4} \\ \frac{P_1 - P_3}{4} & \left(P_1 - \frac{P_3}{2}\right) - \left(\frac{P_2 - P_4}{2}\right) \end{bmatrix}$$

Solving simultaneously:

$P_1 = \frac{P_1}{4} + 4$

$P_2 = \frac{P_1}{2} - \frac{P_3}{4} = \frac{8}{3} - \frac{P_3}{4}$

$P_3 = \frac{P_1}{2} - \frac{P_3}{4} = \frac{8}{3} - \frac{P_3}{4}$

$\frac{3P_1}{4} = 4$

$\frac{5P_2}{4} = \frac{8}{3}$

$P_3 = P_2 = \frac{32}{15}$ \rightsquigarrow symmetric

$P_1 = \frac{16}{3}$

$P_2 = \frac{32}{15}$

$P_4 = P_1 - \frac{P_3}{2} - \frac{P_2}{2} + \frac{P_4}{4} = \frac{16}{3} - \frac{32}{15} + \frac{P_4}{4} = \frac{80-32}{15} + \frac{P_4}{4}$

$\frac{3P_4}{4} = \frac{48}{15} = \frac{16}{5}$

$P_4 = \frac{64}{15}$

2.2-2 contd a. Thus $P = \begin{bmatrix} \frac{16}{3} & \frac{32}{15} \\ \frac{32}{15} & \frac{64}{15} \end{bmatrix}$

b. Since $|\text{eig}(A)| < 1$, A is stable, and since the observability matrix is,

$$\begin{bmatrix} C \\ CA \end{bmatrix} = \begin{bmatrix} 2 & 0 \\ 1 & 2 \end{bmatrix},$$

which has full rank, the solution to the Riccati equation has the unique solution found in part a.

$$2.2-4 \quad \text{Let } X_{k+1} = 2X_k + U_k$$

a. find the homogeneous solution X_k for $k=0, 5$ if $X_0=3$

homogeneous solution $\Rightarrow U_k=0 \forall k$

$$\Rightarrow X_{k+1} = 2X_k$$

K	X_k	X_{k+1}
0	3	6
1	6	12
2	12	24
3	24	48
4	48	96
5	96	192

b. Minimum-energy control to drive state from $X_0=3$ to $X_5=0$

$$\text{cost function: } J_0 = \frac{1}{2} \sum_{k=0}^5 r U_k^2$$

Thus the Hamiltonian is:

$$H^k = \frac{1}{2} r U_k^2 + \lambda_{k+1} (2X_k + U_k)$$

From this we get the state & costate equations:

$$\text{state: } X_{k+1} = \frac{\partial H^k}{\partial \lambda_{k+1}} = 2X_k + U_k$$

$$\text{costate: } \lambda_k = \frac{\partial H^k}{\partial X_k} = 2\lambda_{k+1}$$

Also have the stationarity condition:

$$0 = \frac{\partial H^k}{\partial U_k} = rU_k + \lambda_{k+1} \Rightarrow U_k = -\frac{\lambda_{k+1}}{r}$$

Substituting this U_k into the state equation:

$$X_{k+1} = 2X_k - \frac{\lambda_{k+1}}{r}$$

2.2-4 cont

b)

Next we write λ_k in terms of λ_N :

$$\lambda_{N-1} = 2 \cdot \lambda_N$$

$$\lambda_{N-2} = 2^2 \lambda_N$$

:

$$\lambda_k = 2^{N-k} \lambda_N = 2^{5-k} \lambda_5$$

$$\text{Thus } \lambda_{k+1} = 2^{N-(k+1)} \lambda_N = 2^{N-k-1} \lambda_N = 2^{4-k} \lambda_5$$

Substituting into state equation:

$$x_{k+1} = 2x_k - \frac{2^{4-k} \lambda_5}{r} \quad (\text{1st order difference equation})$$

$$\Rightarrow x_k = 2^k x_0 - \sum_{i=0}^{k-1} 2^{k-i-1} \frac{1}{r} \cdot 2^{N-i-1} \cdot \lambda_N$$

$$x_k = 2^k x_0 - \frac{\lambda_5}{r} \sum_{i=0}^{k-1} 2^{N+k-2i-2}$$

Setting $k=N=5$, we can solve for $\lambda_N = \lambda_5$

$$\lambda_5 = 2^5 (3) - \frac{\lambda_5}{r} \sum_{i=0}^4 2^{5+5-2i-2} = 2^5 \cdot 3 - \frac{\lambda_5}{r} \sum_{i=0}^4 2^{8-2i} = 96 - \frac{\lambda_5}{r} \cdot 341 = 0$$

$$\Rightarrow \lambda_5 = \frac{96}{341} \cdot r$$

$$\text{Thus, } \lambda_{k+1} = 2^{4-k} \cdot \frac{96}{341} r$$

$$\Rightarrow u_k = -\frac{\lambda_{k+1}}{r} = -\frac{2^{4-k} \cdot \frac{96}{341} r}{r} = -2^{4-k} \frac{96}{341}$$

K	u_k	x_k
0	$-2^4 \cdot \frac{96}{341} = 4.5044$	3
1	$-2^3 \cdot \frac{96}{341} = 2.522$	1.4956
2	1.1261	0.7390
3	0.5630	0.3519
4	0.2815	0.1408
5	—	$-9.38 \times 10^{-5} \approx 0$

→ rounding error from MATLAB

2.2-4 (cont'd) C. optimal K_k to minimize $J_0 = \sum_{k=0}^4 (x_k^2 + u_k^2)$

The key is finding the relation:

$$\lambda_N = S_N x_N, \quad S_N = 10$$

To do this we let $\lambda_k = S_k x_k \quad \forall k=0, \dots, N$

To find the sequence of S_k , we solve the Riccati equation backwards:

$$S_k = \frac{a^2 r S_{k+1}}{b^2 S_{k+1} + r} + q$$

This can then be used to find the optimal gains K_k :

$$K_k = \frac{ab S_{k+1}}{b^2 S_{k+1} + r}$$

For our system,

K	S_k	K_k
5	10	—
4	$\frac{4 \cdot 1 \cdot 10}{1 \cdot 10 + 1} + 1 = 4.6364$	$\frac{2 \cdot 10}{10 + 1} = 1.818$
3	4.2903	1.6452
2	4.2439	1.6220
1	4.2372	1.6186
0	4.2362	1.6181

With these gains, we can get the trajectory and cost:

$$U_k = -K_k x_k$$

$$X_{k+1} = 2X_k - K_k x_k = (2 - K_k) x_k, \quad J_k^* = \frac{1}{2} S_k x_k^2$$

K	X_k	J_k^*
0	3	19.0631
1	0.5455	4.74
2	0.1935	1.16
3	0.0732	0.27
4	0.0279	0.046
5	0.0107	$4.4 \times 10^{-8} \sim 0$

$$\underbrace{f}_\phi$$

2.2-7

Let $x_{k+1} = \underbrace{ax_k + bu_k}_{f^k}$ (scalar system)

$$J = \frac{1}{3} \underbrace{\sum_{N=0}^N x_N^3}_{\phi} + \frac{1}{3} \sum_{k=0}^{N-1} (\underbrace{q_f x_k^3 + r u_k^3}_{L^k})$$

a.) Defining the hamiltonian:

$$H^k = L^k + \lambda_{k+1}^f = \frac{1}{3}(q_f x_k^3 + r u_k^3) + \lambda_{k+1}(ax_k + bu_k)$$

State equation: $x_{k+1} = \frac{\partial H^k}{\partial \lambda_{k+1}} = ax_k + bu_k$

Costate equation: $\lambda_k = \frac{\partial H^k}{\partial x^k} = x_k^2 q_f + a \lambda_{k+1}$

Stationarity condition: $0 = \frac{\partial H^k}{\partial u_k} = r u_k^2 + b \lambda_{k+1}$

b. Solving the stationarity condition for u_k :

$$r u_k^2 = -b \lambda_{k+1}$$

$$u_k = \sqrt{-\frac{b}{r} \lambda_{k+1}}$$

\Rightarrow Since we assume $r > 0, \lambda_k > 0$, b must be negative to solve for u_k

Substituting u_k into state (assuming $b < 0$)

$$x_{k+1} = ax_k + b \sqrt{-\frac{b}{r} \lambda_{k+1}}$$

2.2-7 contd C. Solve open-loop problem ($\dot{x}_N = 0$, $s_N = 0$, $q_f = 0$)

The performance index becomes:

$$J = \frac{1}{3} \sum_{k=0}^{N-1} r u_k^3$$

and the Hamiltonian:

$$H^k = \frac{1}{3} r u_k^3 + \lambda_{k+1} (ax_k + bu_k)$$

$$\Rightarrow x_{k+1} = ax_k + bu_k \quad \text{--- state}$$

$$\lambda_k = \frac{\partial H^k}{\partial x_k} = a\lambda_{k+1} \quad \text{--- costate}$$

$$0 = \frac{\partial H^k}{\partial u_k} = r u_k^2 + b\lambda_{k+1} \quad \text{--- stationarity}$$

$$\text{From part b. : } x_{k+1} = ax_k + b\sqrt{-\frac{b}{r}\lambda_{k+1}}$$

Since the costate is a difference equation, we can write

$$\lambda_k = a^{N-k} \lambda_N$$

$$\Rightarrow x_{k+1} = ax_k + b\sqrt{-\frac{b}{r}} a^{\underbrace{N-k-1}_{f}} \lambda_N$$

This new state equation is also a difference equation, with forcing function f .

$$\Rightarrow x_k = a^k x_0 + \sum_{i=0}^{k-1} a^{k-i-1} \left(b\sqrt{-\frac{b}{r}} (a^{N-k-i} \lambda_N) \right)$$

$$= a^k x_0 + b \cdot \sqrt{-\frac{b}{r}} \cdot \sqrt{\lambda_N} \cdot \underbrace{\sum_{i=0}^{k-1} a^{k-i-1} \cdot a^{\frac{(N-k-i)/2}{2}}}_{a^{\frac{N+k-3i-2}{2}}} = a^{\frac{N+k-2}{2}} \cdot \frac{-3i}{a^{\frac{3}{2}}}$$

$$\Rightarrow x_k = a^k x_0 + b \sqrt{-\frac{b}{r}} \lambda_N a^{\frac{N+k-2}{2}} \cdot \sum_{i=0}^{k-1} \left(\frac{1}{a^{\frac{3}{2}}} \right)^i \quad \text{geometric series}$$

$$= a^k x_0 + b \sqrt{-\frac{b}{r}} \lambda_N a^{\frac{N+k-2}{2}} \cdot \frac{1 - a^{-\frac{3}{2}(k)}}{1 - a^{-\frac{3}{2}}} \rightarrow \text{only unknown is } \lambda_N$$

c)

2.2-7_{contd} Let $k=N$

$$x_N = a^N x_0 + b \sqrt{-\frac{b}{r}} \lambda_N a^{N+N-2} \cdot \frac{1 - a^{-\frac{3}{2}N}}{1 - a^{-3/2}}$$

$$\frac{x_N - a^N x_0}{b} = \left(-\frac{b}{r} \lambda_N \right)^{\frac{1}{2}} \left(a^{\frac{N-1}{2N-2}} \right)^{\frac{1}{2}} \frac{1 - a^{-\frac{3}{2}N}}{1 - a^{-\frac{3}{2}}}$$

$$\frac{a^{\frac{2N}{2}} \cdot 1 - a^{-\frac{3N}{2}}}{a^{\frac{3}{2}} \cdot 1 - a^{-\frac{3}{2}}} = \frac{a^{\frac{N}{2}} - a^{-\frac{N}{2}}}{a - a^{-\frac{1}{2}}}.$$

$$\Rightarrow \left(-\frac{b}{r} \lambda_N \right)^{\frac{1}{2}} = \left[\frac{x_N - a^N x_0}{b} \right] \left[\frac{a - a^{\frac{1}{2}}}{a^{\frac{N}{2}} - a^{-\frac{N}{2}}} \right]$$

$$\lambda_N = -\frac{r}{b} \left[\frac{x_N - a^N x_0}{b} \right]^2 \left[\frac{a - a^{\frac{1}{2}}}{a^{\frac{N}{2}} - a^{-\frac{N}{2}}} \right]^2$$

Thus the costate for arbitrary $k \leq N$

$$\lambda_k = -a^{N-k} \frac{r}{b} \cdot \left[\frac{x_N - a^N x_0}{b} \right]^2 \left[\frac{a - a^{\frac{1}{2}}}{a^{\frac{N}{2}} - a^{-\frac{N}{2}}} \right]^2$$

Then the optimal control sequence is:

$$u_k^* = \sqrt{-\frac{b}{r} \lambda_{k+1}} = \left\{ + \frac{b}{r} \frac{r}{b} a^{N-k-1} \cdot \left[\frac{x_N - a^N x_0}{b} \right]^2 \left[\frac{a - a^{\frac{1}{2}}}{a^{\frac{N}{2}} - a^{-\frac{N}{2}}} \right]^2 \right\}^{\frac{1}{2}}$$

$$u_k^* = a^{\frac{N-k-1}{2}} \cdot \frac{x_N - a^N x_0}{b} \cdot \frac{a - a^{\frac{1}{2}}}{a^{\frac{N}{2}} - a^{-\frac{N}{2}}}$$