

Solution to Homework 3 Mathematical Systems Theory, SF2832 Fall 2018 You may use min(5,(your score)/5) as bonus credit on the exam.

1. Consider

$$\dot{x} = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} x + \begin{bmatrix} 0 \\ 1 \end{bmatrix} u$$
$$y = \begin{bmatrix} 2 & 1 \end{bmatrix} x.$$

Answer: omitted.

2. Consider

 $\dot{x}_1 = x_2$ $\dot{x}_2 = \gamma u$ $x(0) = x_0,$

where $\gamma > 0$ can be viewed as a control gain.

(a) Design a stabilizing feedback control $u = k_1 x + k_2 x_2$ that is also the optimal control to

$$\min_u \int_0^\infty (x_1^2(t) + u^2(t))dt.$$

Answer: By solving the ARE, we have $u^* = x_1 - \sqrt{\frac{2}{\gamma}}x_2$.

- (b) What is the optimal cost as $\gamma \to \infty$?.....(2p) Answer: 0.
- **3.** Consider

 $\dot{x}_1 = x_2$ $\dot{x}_2 = ax_1 + u,$

and the cost function

$$J = \int_0^{t_1} (x_2^2 + u^2) dt,$$

where a is a constant.

- 4. Let $P(t, t_f)$ denote the solution at time t with terminal time t_f to the following matrix Riccati equation:

$$\dot{P} = -A^T P - PA + PBB^T P - Q$$
$$P(t_f, t_f) = 0,$$

where A, B, Q are constant matrices and $Q \ge 0$. Show that $P(t_1, t_f) - P(t_2, t_f) \ge 0$ for any $t_1 \le t_2 \le t_f$(3p) **Answer:** When $t_0 < T_1 \le T_2$, we have $x_0^T P(t_0, T_1) x_0 = \min_u \int_{t_0}^{T_1} (x^T Q x + u^T u) dt \le \min_u \int_{t_0}^{T_2} (x^T Q x + u^T u) dt = x_0^T P(t_0, T_2) x_0$, namely $P(t_0, T_1) \le P(t_0, T_2)$. However, $P(t_0, T) = P(0, T - t_0)$, the rest follows if we identify t_f as T and t_1, t_2 as t_0 in the last equality.

5. Consider the discrete time system

$$\begin{split} x(t+1) &= Ax(t) + Bv(t), \\ y(t) &= Cx(t) + Dw(t), \end{split}$$

where

and where v, w, x(0) are uncorrelated with zero mean value.

 Answer: The covariance matrix $V(t) = Ex(t)x(t)^T$ has the initial condition $V(0) = P_0$, and it evolves over time as

$$V(t+1) = Ex(t+1)x(t+1)^T = E\{(Ax(t) + Bv(t))(x(t)^T A^T + v(t)^T B^T)\}$$

= $AV(t)A^T + BB^T$.

(b) Let $\hat{x}(t) = Ex(t)$ be the Kalman estimate defined by

$$\hat{x}(t+1) = A\hat{x}(t) + AK(t)(y(t) - C\hat{x}(t))$$

and let $K(t) = P(t)C^T[CP(t)C^T + DD^T]^{-1}$ be the Kalman gain. Determine a recursive equation for the covariance matrix $P(t) = E\{(x(t) - \hat{x}(t))(x(t) - \hat{x}(t))^T\}$ of the estimation error, where we choose $P(0) = P_0$(1p) **Answer:** $P(t+1) = AP(t)A^T + BB^T - AP(t)C^T[CP(t)C^T + DD^T]^{-1}CP(t)A^T$.

(c) Show that R(t) := V(t) - P(t) is positive semi-definite for all t(3p) **Answer:** We prove it inductively. We have $R(0) = V(0) - P(0) = P_0 - P_0 = 0$, which is positive semidefinite. Now assume $R(t) \ge 0$ and try to show that $R(t+1) \ge 0$.

$$R(t+1) = V(t+1) - P(t+1) = A \underbrace{(V(t) - P(t))}_{R(t) \ge 0} A^{T} + \underbrace{AP(t)C^{T}[CP(t)C^{T} + DD^{T}]^{-1}CP(t)A^{T}}_{>0}$$

so R(t+1) is positive semidefinite. R(t) is now positive semi-definite for all t by induction.