

## A Appendix: Proofs of intermediate results of Section 3

### A.1 Proof of Lemma 1

*Proof.* We start to show that there is a one-to-one correspondence between the positive definite solutions of the discrete algebraic Riccati Equation

$$P = FPF^T - FPH_w^T(H_wPH_w^T + R_w)^{-1}H_wPF^T + LL^T, \quad (14)$$

and of its counterpart in information form

$$M = (FM^{-1}F^T + LL^T)^{-1} + H_w^TR_w^{-1}H_w. \quad (15)$$

Let  $M$  be a positive definite solution of (15), and set  $P := FM^{-1}F^T + LL^T$ . Note that this matrix is positive definite, because it can be written as

$$P = \begin{pmatrix} F & L \end{pmatrix} \begin{pmatrix} M^{-1} & \\ & \mathbf{I}_\ell \end{pmatrix} \begin{pmatrix} F & L \end{pmatrix}^T$$

and our controllability assumption (A1) implies that the matrix  $[F, L]$  has full row rank. ( Indeed, If  $C = [L, FL, F^2L, \dots, F^{n-1}L]$  has rank  $n$  then  $CC^T$  has also rank  $n$ , but  $CC^T = [F, L]B$ , where  $B$  is some matrix. Thus, the  $n \times (n + \ell)$ -matrix  $[F, L]$  must have full rank  $n$ . ) So we can write  $M = P^{-1} + H_w^TR_w^{-1}H_w$ . By the Woodbury matrix identity we have  $M^{-1} = P - PH_w^T(H_wPH_w^T + R_w)^{-1}H_wP$ , and it is now clear that  $P$  solves (14).

Conversely, if  $P$  is a positive solution of Eq (14), set  $M := P^{-1} + H_w^TR_w^{-1}H_w$ , which is also a positive definite matrix. As above, we can write  $M^{-1} = P - PH_w^T(H_wPH_w^T + R_w)^{-1}H_wP$ , and so we have  $P = FM^{-1}F^T + LL^T$ . Finally, we have  $M := P^{-1} + H_w^TR_w^{-1}H_w = (FM^{-1}F^T + LL^T)^{-1} + H_w^TR_w^{-1}H_w$ , i.e.  $M$  solves (15).

Moreover, the two maps that we have defined above are inverses of each other, between the set of positive definite solutions of the DARE and of the I-DARE. Indeed, the map  $M \mapsto FM^{-1}F^T + LL^T$  can be rewritten as  $M \mapsto (M - H_w^TR_w^{-1}H_w)^{-1}$  when  $M$  is a solution of the I-DARE, which is clearly the inverse mapping of  $P \mapsto P^{-1} + H_w^TR_w^{-1}H_w$ .

Then, we can apply known results about the former Riccati equation (14): under our assumption (A1) that  $(F, L)$  is controllable, there exists a positive semidefinite solution to Eq. (14) if and only if the pair  $(F, H_w)$  is detectable, and moreover this solution is unique and positive definite (see e.g. Theorems 23 and 25 in [Sim06]). So the proof of the lemma is complete.

### A.2 Proof of Proposition 2

Let  $\mathbf{w} \in \mathcal{W}$ ,  $X \in \mathcal{X}(\mathbf{w})$  and  $U$  be a matrix satisfying (i) and (ii) in the definition (10) of  $\mathcal{X}(\mathbf{w})$ . Define  $M(\mathbf{w}) := \sum_{i=1}^s \mu_i(w_i) H_i^T H_i$ . In particular,

$$\begin{pmatrix} U + M(\mathbf{w}) - F^T U F & F^T U L \\ L^T U F & \mathbf{I}_\ell - L^T U L \end{pmatrix} \succeq 0.$$

Since  $(F, L)$  is controllable, there is a gain matrix  $N$  such that  $F + LN$  has all his eigenvalues outside the unit disc of  $\mathbb{C}$ , see [KSH00, Sections C3 and C4], and hence every eigenvalue of  $\Gamma = (F + LN)^{-1}$  lies inside the open unit disc. Now, we obtain another positive semidefinite matrix by a congruence operation:

$$\begin{pmatrix} \mathbf{I}_n & -N^T \end{pmatrix} \begin{pmatrix} U + M(\mathbf{w}) - F^T U F & F^T U L \\ L^T U F & \mathbf{I}_\ell - L^T U L \end{pmatrix} \begin{pmatrix} \mathbf{I}_n \\ -N \end{pmatrix} \succeq 0.$$

This gives:

$$U + M(\mathbf{w}) + N^T N - (F + LN)^T U (F + LN) \succeq 0.$$

Pre and post-multiplying by  $\Gamma^T$  (resp.  $\Gamma$ ), we obtain

$$U \preceq \Gamma^T U \Gamma + Q_{\mathbf{w}},$$

where  $Q_{\mathbf{w}} := \Gamma^T (M(\mathbf{w}) + N^T N) \Gamma \succeq 0$ . Denote by  $U_0$  the unique solution of the discrete-time Lyapunov equation  $U = \Gamma^T U \Gamma + Q_{\mathbf{w}}$ , see [KSH00, Section D1]. We have

$$U - U_0 \preceq \Gamma^T (U - U_0) \Gamma \preceq \dots \preceq (\Gamma^T)^k (U - U_0) \Gamma^k \preceq \dots,$$

and so  $U \preceq U_0$  because  $\Gamma^k$  converges to 0 as  $k \rightarrow \infty$ .

Now, observe that  $U_0$  has a closed-form expression that can be obtained by vectorizing the Lyapunov equation:  $\text{vec}(U_0) = (\mathbf{I}_{n^2} - \Gamma^T \otimes \Gamma^T)^{-1} \text{vec}(Q_{\mathbf{w}})$ , where  $\otimes$  denotes the Kronecker product. Using this relation, we obtain  $\|U_0\|_F \leq \|(\mathbf{I}_{n^2} - \Gamma^T \otimes \Gamma^T)^{-1}\|_2 \cdot \|Q_{\mathbf{w}}\|_F$ , where  $\|M\|_F := \sqrt{\text{trace}(MM^T)}$  denotes the Frobenius norm of  $M$  and  $\|M\|_2$  is its spectral norm. By using the definition of  $Q_{\mathbf{w}}$ , we can thus conclude that there exists a constant  $\alpha' \geq 0$  such that  $\|U\|_F \leq \alpha' (1 + \sum_i \mu_i(w_i) \|H_i\|_F^2)$ . Finally, we obtain the bound of the proposition,  $\|X\|_2 \leq \alpha (1 + \sum_i \mu_i(w_i) \|H_i\|_2^2)$  for some  $\alpha \geq 0$ , by using (i) in Eq. (10) and the fact that the Frobenius and spectral norms are equivalent.

### A.3 Proof of Proposition 3

Let  $\mathbf{w} \in \mathcal{W}$ . We introduce the function  $g$  that maps  $\mathbb{S}_n^{++}$  onto  $\mathbb{S}_n$ , defined by  $g(X) = f(X, \mathbf{w}) = (FX^{-1}F^T + LL^T)^{-1} + M(\mathbf{w}) - X$ . Note that  $X \in \mathcal{X}^+(\mathbf{w})$  if and only if  $g(X) \succeq 0$ .

The directional derivative of  $g$  in the direction of  $\Delta \succeq 0$  can be found by using the formula  $\frac{dA^{-1}}{dt} = -A^{-1} \frac{dA}{dt} A^{-1}$  and the chain rule:

$$\begin{aligned} Dg(M)[\Delta] &:= \lim_{\varepsilon \rightarrow 0} \frac{1}{\varepsilon} (g(M + \varepsilon \Delta) - g(M)) \\ &= (FM^{-1}F^T + LL^T)^{-1} FM^{-1} \Delta M^{-1} F^T (FM^{-1}F^T + LL^T)^{-1} - \Delta. \end{aligned}$$

Let  $X$  be a matrix in  $\mathcal{X}^+(\mathbf{w})$ , such that  $g(X) \neq 0$ . We are going to show that there exists a matrix  $X'$  in the neighborhood of  $X$  satisfying  $X' \succeq X$ ,  $X' \neq X$ ,  $g(X') \succeq 0$ . This is equivalent to the following first order property (see e.g., [HUL93, Section VI.5.1]):

$$\exists \Delta \succeq 0 : \Delta \neq 0, \quad \forall \mathbf{u} \in \text{Ker}(g(X)), \quad \mathbf{u}^T Dg(X)[\Delta] \mathbf{u} \geq 0.$$

This condition is satisfied for  $\Delta := g(X)$  indeed, because the first term of  $Dg(X)[\Delta]$  is a positive semidefinite matrix, and the term  $-\mathbf{u}^T \Delta \mathbf{u}$  vanishes for all  $\mathbf{u} \in \text{Ker}(g(X))$ . Hence, for all  $X_0 \in \mathcal{X}^+(\mathbf{w})$  we can define a sequence of matrices  $X_i \in \mathcal{X}^+(\mathbf{w})$  satisfying  $X_0 \preceq X_1 \preceq \dots X_n \preceq \dots$  as follows:  $X_{n+1} = \Psi(X_n)$ , where

$$\Psi(X) = X + \left( \operatorname{argmax}_{t \geq 0} \left\{ t : g(X + t g(X)) \succeq 0 \right\} \right) g(X).$$

The sequence  $(X_i)$  is bounded (by Propositions 1 and 2) and hence it converges. (Indeed, if it has two accumulation points  $X_1^*$  and  $X_2^*$ , then for all  $\mathbf{u} \in \mathbb{R}^n$  the non-decreasing sequence  $\mathbf{u}^T X_i \mathbf{u}$  converges to some value, which must be equal to both  $\mathbf{u}^T X_1^* \mathbf{u}$  and  $\mathbf{u}^T X_2^* \mathbf{u}$ . Hence,  $\mathbf{u}^T (X_1^* - X_2^*) \mathbf{u} = 0$  for all  $\mathbf{u}$ , which proves  $X_1^* = X_2^*$ .) Denote this limit by  $X^*$ . The above discussion shows that  $g(X^*) = 0$ , otherwise  $X^*$  cannot be a fixed point of  $\Psi$ . This means that  $X^*$  is the unique positive definite solution of the I-DARE (3). We have thus  $X_0 \preceq X^* = M_\infty(\mathbf{w})$ , and the proposition is proved.

## References

- [HUL93] J.B. Hiriart-Urruty and C. Lemaréchal. *Convex analysis and minimization algorithms: Fundamentals*. Springer, 1993.
- [KSH00] T. Kailath, A.H. Sayed, and B. Hassibi. *Linear estimation*. 2000.
- [Sim06] D. Simon. *Optimal state estimation: Kalman,  $H_\infty$ , and nonlinear approaches*. Wiley, 2006.