

Guaranteed Cost Bounds for Robust Stability and Performance Analysis of Discrete-Time Systems

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Abstract

In this paper we derive new guaranteed cost bounds for robust stability and performance with real structured uncertainty for discrete-time systems. In particular, we obtain a shifted bounded real bound, a linear bound, a shifted linear bound, an inverse bound, a shifted inverse bound, and a shifted Popov bound. Several examples are used to compare these new bounds.

1. Introduction

The analysis and synthesis of robust controller has been of intense interest during the past three decades [1]. While unstructured complex-valued uncertainty can be addressed nonconservatively using quadratic bounds [2], structured real-valued uncertainty is a more difficult problem due to the discontinuity of the structured singular value and stability margins [3]. The most effective approach to this problem has been the development of frequency-dependent scales and multipliers that account for phase restrictions on the parametric or dynamic uncertainty [4, 5].

For stability analysis with real polytopic uncertainty, LMI techniques can be used to solve multiple Lyapunov equations to determine stability bounds [6]. This approach avoids the need for frequency-dependent multipliers and utilizes convex optimization methods to obtain common Lyapunov functions.

An alternative approach that is applicable to controller synthesis is Riccati-based methods which provide guaranteed cost bounds for the worst-case H_2 performance. For continuous-time systems a large class of guaranteed cost bounds have been developed. While the small gain bounds are the best known due to their connections with the small gain theorem [7, 8], alternative bounds have been developed as well [9]-[13]. For structured real-valued uncertainty, these bounds have significantly reduced conservatism as compared to small gain bounds.

For discrete-time systems quadratic bounds have been developed in the context of bounded-real theory (see [14] and the references given therein). Discrete-time Popov bounds given in [15, 16] also provide quadratic tests for robust stability and performance. For a sampled-data system with parametric uncertainty, a nonquadratic bound was developed in [17]. Compared to continuous-time systems, however, there has been relatively little effort devoted to the development of discrete-time bounds.

The objective of the present paper is to develop novel bounds for structured real uncertainty for discrete-time systems. Some of these bounds can be viewed as the counterpart of bounds developed for continuous-time systems in [13]. Unlike [17] we consider general uncertainty structures rather than sampled-data uncertainty structures. We consider parameter-independent bounds, each equivalent to a common Lyapunov function, and parameter-dependent bounds which are equivalent to multiple Lyapunov functions.

In Section 2 we present the main robustness result that provides the basis for specific bounds given later in the paper. In Section 3 we present an LMI approach to bounding polytopic uncertainties. In Sections 4, 5, 6, and 7, we present the discrete-time forms of the shifted bounded real bound, the linear and shifted linear bounds, the inverse and shifted inverse bounds, and the shifted

Popov Bound. Finally, in Section 8, we present several numerical examples to compare the different bounds.

Notation

\mathbf{R}^d	$d \times 1$ real column vectors
$\mathbf{R}^{m \times n}$	$m \times n$ real matrices
I_n, \mathbf{N}^n	$n \times n$ identity matrix, nonnegative-definite matrices
$\mathbf{S}^n, \mathbf{P}^n$	$n \times n$ symmetric matrices, positive-definite matrices
$A \leq B$	$B - A$ is nonnegative definite
$\mathcal{E}(\cdot)$	expectation operator
tr	trace operator
$ H $	$(HH^T)^{\frac{1}{2}}$, where $H \in \mathbf{R}^{n \times n}$

2. Robust Performance and Guaranteed Cost Bounds

Let $\mathcal{U} \subset \mathbf{R}^{n \times n}$ denote an uncertainty set and consider the discrete-time system

$$x(k+1) = (A + \Delta A)x(k) + Dw(k), \quad (1)$$

where $x \in \mathbf{R}^n$ and $w \in \mathbf{R}^d$ are the state and disturbance, respectively, $\Delta A \in \mathcal{U}$, and the disturbance w is a standard zero-mean white noise process. We assume throughout that A is asymptotically stable. Next, consider the performance variable

$$z(k) = Ex(k). \quad (2)$$

If $A + \Delta A$ is asymptotically stable for all $\Delta A \in \mathcal{U}$, then define the worst-case H_2 performance measure

$$J(\mathcal{U}) = \sup_{\Delta A \in \mathcal{U}} \text{tr } P_{\Delta A} V, \quad (3)$$

where $V \triangleq DD^T$ and $P_{\Delta A}$ is the nonnegative-definite solution to the Lyapunov equation

$$P_{\Delta A} = (A + \Delta A)^T P_{\Delta A} (A + \Delta A) + R, \quad (4)$$

where $R \triangleq E^T E$.

The following result, which is an extension of Theorem 3.1 of [18], provides a bound for the worst-case cost $J(\mathcal{U})$.

Theorem 1 Let $\Omega : \mathcal{N} \subseteq \mathbf{S}^n \rightarrow \mathbf{S}^n$ be such that there exists $P \in \mathcal{N}$ satisfying

$$P = A^T P A + \Omega(P) + R, \quad (5)$$

and let $P_0 : \mathcal{U} \rightarrow \mathbf{S}^n$ be such that

$$0 \leq P + P_0(\Delta A), \quad \Delta A \in \mathcal{U}, \quad (6)$$

and

$$0 \leq \Omega(P) - [(A + \Delta A)^T P_0(\Delta A)(A + \Delta A) - P_0(\Delta A)] - [\Delta A^T P A + A^T P \Delta A + \Delta A^T P \Delta A], \quad \Delta A \in \mathcal{U}. \quad (7)$$

Then $(A + \Delta A, E)$ is detectable for all $\Delta A \in \mathcal{U}$ if and only if $A + \Delta A$ is asymptotically stable for all $\Delta A \in \mathcal{U}$. In this case,

$$P_{\Delta A} \leq P + P_0(\Delta A), \quad \Delta A \in \mathcal{U}, \quad (8)$$

where $P_{\Delta A} \in \mathbf{N}^n$ is given by (4), and

$$J(\mathcal{U}) \leq \text{tr } P V + \sup_{\Delta A \in \mathcal{U}} \text{tr } P_0(\Delta A) V. \quad (9)$$

If, in addition, there exists $\bar{P}_0 \in \mathbf{S}^n$ such that

$$P_0(\Delta A) \leq \bar{P}_0, \quad \Delta A \in \mathcal{U}, \quad (10)$$

then

$$J(\mathcal{U}) \leq \text{tr}[(P + \bar{P}_0)V]. \quad (11)$$

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The pair (Ω, P_0) is a *bound*. A bound is *parameter-independent* if P_0 is constant. In this case, we write P_0 for $P_0(\Delta A)$ and set $\tilde{P}_0 = P_0$. Finally, a bound is *parameter-dependent* if P_0 depends on ΔA . For a given bound (Ω, P_0) , the following result yields an equivalent bound $(\hat{\Omega}, \hat{P}_0)$.

Proposition 1 Let $\Omega : \mathcal{N} \subseteq \mathbf{S}^n \rightarrow \mathbf{S}^n$, $P \in \mathcal{N}$, and $P_0 : \mathcal{U} \rightarrow \mathbf{S}^n$ satisfy (5)-(7), and let $\tilde{P}_0 \in \mathbf{S}^n$ satisfy (10). Let $\hat{P}_0 \in \mathbf{S}^n$, and define $\hat{\mathcal{N}} \subseteq \mathbf{S}^n$, $\hat{\Omega} : \hat{\mathcal{N}} \rightarrow \mathbf{S}^n$ and $\hat{P}_0 : \mathcal{U} \rightarrow \mathbf{S}^n$ by $\hat{\mathcal{N}} \triangleq \mathcal{N} + \tilde{P}_0 - \hat{P}_0$,

$$\hat{\Omega}(P + \tilde{P}_0 - \hat{P}_0) \triangleq \Omega(P) - A^T(\tilde{P}_0 - \hat{P}_0)A + (\tilde{P}_0 - \hat{P}_0), \quad (12)$$

and

$$\hat{P}_0(\Delta A) \triangleq P_0(\Delta A) - \tilde{P}_0 + \hat{P}_0. \quad (13)$$

Then Theorem 1 is satisfied with \mathcal{N} , Ω , P , P_0 , and \tilde{P}_0 replaced by $\hat{\mathcal{N}}$, $\hat{\Omega}$, $P + \tilde{P}_0 - \hat{P}_0$, \hat{P}_0 , and \hat{P}_0 .

Remark 1 In Proposition 1, if $P_0(\Delta A) = P_0 = \tilde{P}_0$, $\Delta A \in \mathcal{U}$, then $\hat{P}_0(\Delta A) = \hat{P}_0 = \tilde{P}_0$, $\Delta A \in \mathcal{U}$. In particular, choosing $\tilde{P}_0 = 0$ implies $\hat{P}_0(\Delta A) = 0$, $\Delta A \in \mathcal{U}$.

3. LMI's for Robust Performance

In this section we use linear matrix inequalities to determine the least conservative parameter-independent bound for robust stability and performance with polytopic uncertainty. Let \mathcal{U} have the form

$$\mathcal{U} = \left\{ \Delta A : \Delta A = \sum_{i=1}^r \delta_i A_i, \text{ where } |\delta_i| \leq \gamma, i = 1, \dots, r \right\}, \quad (14)$$

where $A_1, \dots, A_r \in \mathbf{R}^{n \times n}$.

Proposition 2 Let $P \in \mathbf{N}^n$ satisfy the 2^r constraints

$$\begin{aligned} (A + \gamma A_1 + \dots + \gamma A_r)^T P (A + \gamma A_1 + \dots + \gamma A_r) - P + R &\leq 0, \\ (A - \gamma A_1 + \dots + \gamma A_r)^T P (A - \gamma A_1 + \dots + \gamma A_r) - P + R &\leq 0, \\ &\vdots \\ (A - \gamma A_1 - \dots + \gamma A_r)^T P (A - \gamma A_1 - \dots + \gamma A_r) - P + R &\leq 0, \\ (A - \gamma A_1 - \dots - \gamma A_r)^T P (A - \gamma A_1 - \dots - \gamma A_r) - P + R &\leq 0. \end{aligned}$$

Then

$$(A + \Delta A)^T P (A + \Delta A) - P + R \leq 0, \Delta A \in \mathcal{U}.$$

Then $(A + \Delta A, E)$ is detectable for all $\Delta A \in \mathcal{U}$ if and only if $A + \Delta A$ is asymptotically stable for all $\Delta A \in \mathcal{U}$. In this case,

$$P_{\Delta A} \leq P, \Delta A \in \mathcal{U}.$$

Remark 2 Minimizing the convex objective $J(\mathcal{U}) = \text{tr } PV$ subject to the LMI's given in Proposition 2, is a convex optimization problem. In addition, the optimal cost $J(\mathcal{U})$ from this optimization provides the lowest possible cost for a parameter-independent bound under polytopic uncertainty.

4. Shifted Bounded Real Bound

Let the uncertainty set \mathcal{U} be given by

$$\mathcal{U} = \{ \Delta A : \Delta A = B_0 F C_0, \text{ where } F \in \mathcal{F} \}, \quad (15)$$

where \mathcal{F} is a subset of

$$\mathcal{F}_{\text{BR}} \triangleq \{ F \in \mathbf{R}^{k_1 \times k_2} : F^T F \leq M \}, \quad (16)$$

where $M \in \mathbf{N}^{k_2}$, $B_0 \in \mathbf{R}^{n \times k_1}$, and $C_0 \in \mathbf{R}^{k_2 \times n}$. The following result concerns the classical *bounded real bound* [19, 20, 15].

Proposition 3 Define $\mathcal{N} = \{ P \in \mathbf{N}^n : B_0^T P B_0 < I \}$. Suppose there exists $P \in \mathcal{N}$ satisfying

$$P = A^T P A + A^T P B_0 (I - B_0^T P B_0)^{-1} B_0^T P A + C_0^T M C_0 + R. \quad (17)$$

Then (6) and (7) are satisfied with $P_0 = 0$.

Next we present a variation of Proposition 3, the *shifted bounded real bound*.

Proposition 4 Let $T \in \mathbf{P}^{k_1}$, $N, H \in \mathbf{R}^{k_1 \times n}$, and $Z \in \mathbf{S}^n$. Define $\mathcal{N} = \{ P \in \mathbf{N}^n : B_0^T P B_0 < T \}$. Suppose there exists $P \in \mathcal{N}$ satisfying

$$P = A_s^T P A_s + (A_s^T P B_0 - H^T) (T - B_0^T P B_0)^{-1} (B_0^T P A_s - H) + Z + R, \quad (18)$$

where

$$\begin{aligned} (N^T T N + H^T N + N^T H) + (N^T T + H^T) F C_0 \\ + C_0^T F^T (T N + H) + C_0^T F^T T F C_0 \leq Z, F \in \mathcal{F}, \end{aligned} \quad (19)$$

and $A_s \triangleq A - B_0 N$. Then (6) and (7) are satisfied with $P_0 = 0$.

Note that the shifted bounded real equation (18) with $H = 0$, $T = I$, $N = 0$ and $Z = C_0^T M C_0$ yields the bounded real bound equation (17). Letting $H = 0$ and $T = I$, we obtain the discrete-time version of the continuous-time shifted bounded real bound [21].

Corollary 1 Define $\mathcal{N} = \{ P \in \mathbf{N}^n : B_0^T P B_0 < I \}$. Suppose there exists $P \in \mathcal{N}$ satisfying

$$P = A_s^T P A_s + A_s^T P B_0 (I - B_0^T P B_0)^{-1} B_0^T P A_s + M_s + R, \quad (20)$$

where $M_s \in \mathbf{N}^n$ satisfies

$$N^T N + N^T F C_0 + C_0^T F^T N + C_0^T F^T F C_0 \leq M_s, F \in \mathcal{F}. \quad (21)$$

Then (6) and (7) are satisfied with $P_0 = 0$.

Next, assume that \mathcal{U} is given by (14), where $A_1, \dots, A_r \in \mathbf{R}^{n \times n}$. Furthermore, let $B_i \in \mathbf{R}^{n \times k_i}$ and $C_i \in \mathbf{R}^{k_i \times n}$ satisfy $A_i = B_i C_i$, $i = 1, \dots, r$, and let

$$B_0 = [B_1 \quad \dots \quad B_r] \in \mathbf{R}^{n \times k}, \quad C_0 = \begin{bmatrix} C_1 \\ \vdots \\ C_r \end{bmatrix} \in \mathbf{R}^{k \times n}, \quad (22)$$

where $k = \sum_{i=1}^r k_i$. Then $\Delta A = B_0 F C_0$, where $F \in \mathcal{F}$, and \mathcal{F} is given by

$$\mathcal{F} = \{ F = \text{block-diag}(\delta_1 I_{k_1}, \dots, \delta_r I_{k_r}) \in \mathbf{R}^{k \times k} \text{ where } \delta_i \in [-\gamma, \gamma] \}, \quad (23)$$

and M in (16) can be chosen to be $\gamma^2 I_k$. Note that \mathcal{F} is a proper subset of \mathcal{F}_{BR} .

One choice for M_s is given by

$$M_s = N^T N + \gamma |N^T C_0 + C_0^T N| + \gamma^2 C_0^T C_0, \quad (24)$$

where $|H| = (H H^T)^{1/2}$. With this choice of M_s , (20) becomes

$$\begin{aligned} P = A_s^T P A_s + A_s^T P B_0 (I - B_0^T P B_0)^{-1} B_0^T P A_s \\ + N^T N + \gamma |N^T C_0 + C_0^T N| + \gamma^2 C_0^T C_0 + R. \end{aligned} \quad (25)$$

Similarly, (19) is satisfied with

$$\begin{aligned} Z = (N^T T N + H^T N + N^T H) \\ + \gamma |(N^T T + H^T) C_0 + C_0^T (T N + H)| + \gamma^2 C_0^T T C_0. \end{aligned}$$

Proposition 5 Suppose $r = 1$, $A = A^T$ is invertible, and $A_1 = -A_1^T$. Furthermore, suppose there exists $m > 0$, such that

$$\frac{1}{m}ARA < A - A[(A + \gamma^2 A_1^T A^{-1} A_1)]A.$$

Then

$$P = m(A + \gamma^2 A_1^T A^{-1} A_1) + R$$

is a solution to (18) where $B_0 = A$, $C_0 = A^{-1} A_1$, $N = 0$, $T = mA$, and $H = T$. Furthermore, if $P \geq 0$ then (6) and (7) are satisfied with $P_0 = 0$.

5. Shifted Linear Bound

In this section, we let the uncertainty set \mathcal{U} be given by (14). The following result provides the discrete-time form of the continuous-time *linear bound* [22]-[24].

Proposition 6 Define $A_0 = \sum_{i=1}^r A_i$. Let $0 < \alpha < 1$, and suppose there exists $P \in \mathbf{N}^n$ satisfying

$$P = \frac{1}{1-\alpha} A^T P A + \frac{\gamma^2}{\alpha} A_0^T P A_0 + R. \quad (26)$$

Then (6) and (7) are satisfied with $P_0 = 0$.

The solution to (26) can be written as

$$P = \text{vec}^{-1} \left(\left[I - \frac{1}{1-\alpha} (A^T \otimes A^T) - \frac{\gamma^2}{\alpha} (A_0^T \otimes A_0^T) \right]^{-1} \text{vec} R \right).$$

Next we obtain the *shifted linear bound*, the discrete-time version of the continuous-time shifted linear bound [13].

Proposition 7 Define $A_0 = \sum_{i=1}^r A_i$. Let $\alpha > 0$, $N \in \mathbf{S}^n$, and define $\mathcal{N} = \{P \in \mathbf{N}^n : (1-\alpha)(P-N) \geq 0\}$. Suppose there exists $P \in \mathcal{N}$ satisfying

$$P = \frac{1}{1-\alpha} A^T P A + \frac{\gamma^2}{\alpha} A_0^T P A_0 - \frac{\alpha}{1-\alpha} A^T N A - \frac{1-\alpha}{\alpha} \gamma^2 A_0^T N A_0 + \gamma \sum_{i=1}^r |A_i^T N A + A^T N A_i| + R. \quad (27)$$

Then (6) and (7) are satisfied with $P_0 = 0$.

6. Shifted Inverse Bound

In this section we let the uncertainty set \mathcal{U} be given by (14). The following result provides a discrete-time form of the *inverse bound* [10].

Proposition 8 Let $\alpha > 0$, suppose A is invertible, and suppose there exists $P \in \mathbf{P}^n$ satisfying

$$P = (1 + \gamma\alpha r) A^T P A + R + \gamma \sum_{i=1}^r [\gamma A_i^T P A_i + \frac{1}{4\alpha} (A_i^T P A + A^T P A_i) (A^T P A)^{-1} (A_i^T P A + A^T P A_i)]. \quad (28)$$

Then (6) and (7) are satisfied with $P_0 = 0$.

Note that (28) can be written as

$$P = (1 + \gamma\alpha r) A^T P A + \sum_{i=1}^r A^T \left[\frac{\gamma}{4\alpha} (\tilde{A}_i^T P + P \tilde{A}_i) \right] \times P^{-1} (\tilde{A}_i^T P + P \tilde{A}_i) + \gamma^2 \tilde{A}_i^T P \tilde{A}_i A + R,$$

where $\tilde{A}_i = A_i A^{-1}$. Next we obtain the *shifted inverse bound*, the discrete-time version of the continuous-time shifted inverse bound [13].

Proposition 9 Let $\alpha > 0$, $M, N \in \mathbf{S}^n$, suppose A is invertible, and suppose there exists $P \in \mathbf{N}^n \cap (\mathbf{P}^n + N)$ satisfying

$$P = (1 + \alpha\gamma) A^T P A + \sum_{i=1}^r A^T (\gamma^2 \tilde{A}_i^T P \tilde{A}_i - \alpha\gamma N + \frac{\gamma}{4\alpha} [\tilde{A}_i^T (P - M) + (P - M) \tilde{A}_i] (P - N)^{-1} \times [\tilde{A}_i^T (P - M) + (P - M) \tilde{A}_i] + \gamma [\tilde{A}_i^T M + M \tilde{A}_i]) A + R. \quad (29)$$

Then (6) and (7) are satisfied with $P_0 = 0$.

If $N = M$ then (29) becomes

$$P = (1 + \alpha\gamma) A^T P A + \sum_{i=1}^r A^T (\gamma^2 \tilde{A}_i^T P \tilde{A}_i - \alpha\gamma N + \frac{\gamma}{4\alpha} [\tilde{A}_i^T (P - N) + (P - N) \tilde{A}_i] (P - N)^{-1} \times [\tilde{A}_i^T (P - N) + (P - N) \tilde{A}_i] + \gamma [\tilde{A}_i^T N + N \tilde{A}_i]) A + R. \quad (30)$$

With $N = M = 0$, the shifted bound (28) is obtained. Alternatively, with $M = 0$, (29) becomes

$$P = (1 + \alpha\gamma) A^T P A + \sum_{i=1}^r A^T (\gamma^2 \tilde{A}_i^T P \tilde{A}_i - \alpha\gamma N + \frac{\gamma}{4\alpha} [\tilde{A}_i^T P + P \tilde{A}_i] (P - N)^{-1} [\tilde{A}_i^T P + P \tilde{A}_i]) A + R. \quad (31)$$

7. Shifted Popov Bound

Define the Popov uncertainty set \mathcal{F}_{Pop} as

$$\mathcal{F}_{\text{Pop}} \triangleq \{F \in \mathbf{S}^k : M_L \leq F \leq M_U\}, \quad (32)$$

where $M_L, M_U \in \mathbf{S}^k$ are such that $M \triangleq M_U - M_L$ is positive definite. The following result concerns the *Popov bound* [15, 16]. Let \mathcal{U} be given by (15), where $\mathcal{F} \subseteq \mathcal{F}_{\text{Pop}}$.

Proposition 10 Let $N \in \mathbf{R}^{k \times k}$ and $\mu \in \mathbf{S}^k$ satisfy

$$N^T (F - M_L) = (F - M_L) N \leq \mu, \quad F \in \mathcal{F}, \quad (33)$$

$$R_0 \triangleq 2M^{-1} - NC_0 B_0 - (NC_0 B_0)^T - B_0^T C_0^T \mu C_0 B_0 - B_0^T P B_0 > 0. \quad (34)$$

Furthermore, suppose there exists $P \in \mathbf{P}^n$ satisfying

$$P = \tilde{A}_{\text{Pop}}^T P \tilde{A}_{\text{Pop}} + (\tilde{A}_{\text{Pop}} - I)^T C_0^T \mu C_0 (\tilde{A}_{\text{Pop}} - I) + R + [C_0 + (NC_0 + B_0^T C_0^T \mu C_0) (\tilde{A}_{\text{Pop}} - I) + B_0^T P \tilde{A}_{\text{Pop}}]^T \times R_0^{-1} [C_0 + (NC_0 + B_0^T C_0^T \mu C_0) (\tilde{A}_{\text{Pop}} - I) + B_0^T P \tilde{A}_{\text{Pop}}], \quad (35)$$

where $\tilde{A}_{\text{Pop}} \triangleq A + B_0 M_L C_0$. Then (7) is satisfied with $P_0 : \mathcal{U} \rightarrow \mathbf{S}^n$ given by

$$P_0(\Delta A) = C_0^T (F - M_L) N C_0. \quad (36)$$

Remark 3 $\tilde{P}_0 = C_0^T \mu C_0$ satisfies (10).

Next we present a *shifted Popov bound*. Let \mathcal{U} be given by (15), where $\mathcal{F} \subseteq \mathcal{F}_{\text{Pop}}$.

Proposition 11 Let $N \in \mathbf{R}^{k \times k}$, $\mu \in \mathbf{S}^k$, $Z \in \mathbf{N}^n$ and $H \in \mathbf{P}^k$ satisfy

$$(F - M_L) N = N^T (F - M_L) \leq \mu, \quad F \in \mathcal{F}, \quad (37)$$

$$H (F - M_L) N = N^T (F - M_L) H, \quad F \in \mathcal{F}, \quad (38)$$

$$H (F - M_L) = (F - M_L) H, \quad F \in \mathcal{F}, \quad (39)$$

$$R_0 \triangleq 2M^{-1} - NC_0 B_0 H^{-1} - (NC_0 B_0 H^{-1})^T - H^{-1} B_0^T C_0^T H \mu C_0 B_0 H^{-1} - H^{-1} B_0^T P B_0 H^{-1} > 0,$$

and

$$\begin{aligned} & \left[B_0 X^T (F - M_L) H C_0 + C_0^T H (F - M_L) X B_0^T \right] \\ & + (A_{\text{Pop}} - I)^T C_0^T H \mu C_0 (A_{\text{Pop}} - I) \leq Z, \quad F \in \mathcal{F}. \end{aligned}$$

Furthermore, suppose there exists $P \in \mathbb{P}^n$ satisfying

$$\begin{aligned} P = & A_{\text{Pop}}^T P A_{\text{Pop}} + \left[H C_0 + (N C_0 + H^{-1} B_0^T C_0^T H \mu C_0) (A_{\text{Pop}} - I) \right. \\ & \left. + H^{-1} B_0^T P A_{\text{Pop}} - X B_0^T \right]^T R_0^{-1} \left[H C_0 + H^{-1} B_0^T P A_{\text{Pop}} \right. \\ & \left. + (N C_0 + H^{-1} B_0^T C_0^T H \mu C_0) (A_{\text{Pop}} - I) - X B_0^T \right] \\ & + R + Z \end{aligned} \quad (40)$$

Then (7) is satisfied with $P_0 : \mathcal{U} \rightarrow \mathbb{S}^n$ given by

$$P_0(\Delta A) = C_0^T H (F - M_L) N C_0. \quad (41)$$

Remark 4 Proposition 11, gives the discrete-time version of the continuous-time shifted Popov bound [13].

Remark 5 Setting $X = 0$, $Z = 0$, and $H = I$ in Proposition 11 yields Proposition 10.

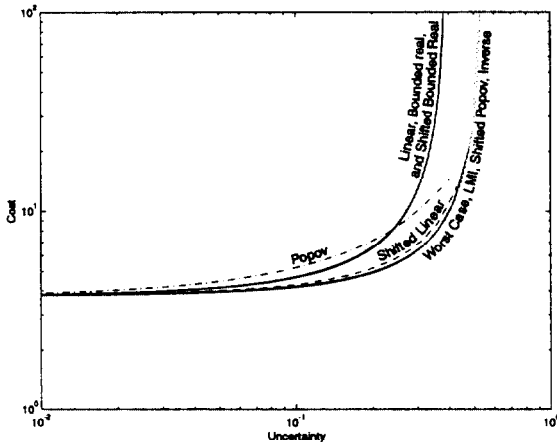


Figure 1: Cost bounds for Example 1.

8. Examples

LMI methods can be used to calculate solutions along with optimal scalings for bounds such as the linear, bounded real, inverse, and Popov bounds, as well as their shifted counterparts. For more on the setup of LMIs for solving bounds, see [13]. In the numerical examples that follow, LMI methods were used to obtain the best parameter-independent bound along with the Popov and shifted Popov bounds.

Example 1 Let

$$A = \begin{bmatrix} 0.523 & -0.307 \\ 0.307 & 0.523 \end{bmatrix}, \quad A_0 = \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix}.$$

Furthermore, let $R = \begin{bmatrix} 0.25 & 0.12 \\ 0.12 & .3 \end{bmatrix}$, and $V = \begin{bmatrix} 4 & 1 \\ 1 & 4 \end{bmatrix}$. Figure 1 shows the actual worst case cost, along with the best possible parameter-independent bound given by Remark 2. Next, let B_0 and C_0 be given by

$$B_0 = \begin{bmatrix} 0 & \varepsilon^{-1} \\ \varepsilon^{-1} & 0 \end{bmatrix}, \quad C_0 = \begin{bmatrix} -\varepsilon & 0 \\ 0 & \varepsilon \end{bmatrix},$$

where $\varepsilon \neq 0$ is a free parameter used for optimization. The bounded real bound, shifted bounded real, linear, and Popov bound,

predict stability for $|\delta| \leq 0.4$. Figure 1 shows the inverse bound and the shifted linear bound, with $N = \beta I$, predicting stability for $|\delta| < 0.45$, with the inverse bound giving a tight fit to the worst case cost in this range. Finally, the shifted Popov bound predicts stability for $|\delta| < 0.48$, with performance coinciding with the actual worst case performance in that region.

Example 2 Consider the problem given in Example 1, with $A = \begin{bmatrix} 0.523 & -0.307 \\ -0.307 & 0.523 \end{bmatrix}$. Figure 2 shows the actual worst case cost, along with the best possible parameter-independent bound given by Remark 2. In this example, the inverse and Popov bounds are conservative predicting stability for $|\delta| \leq 0.035$ and $|\delta| < 0.11$, respectively. The bounded real bound and linear bound predict stability for $|\delta| < 0.17$ and $|\delta| < 0.27$ respectively. The shifted bounded real bound given in Proposition 5 performs slightly better than the linear bound giving better performance for $|\delta| < 0.28$. The shifted linear bound with $N = \beta I$ extends the predicted stability range to $|\delta| < 0.4$, which falls short of the LMI bound prediction of $|\delta| < 0.45$. In contrast, the shifted Popov bound predicts stability beyond the LMI bound, up to $|\delta| < 0.48$.

Example 3 Finally, consider the problem given in Example 2 with $A_0 = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$. Figure 3 shows the actual worst case cost, along with the best possible parameter-independent bound given by Remark 2. The inverse bound predicts stability for $|\delta| \leq 0.07$ and shifted inverse bound with $M = 0$ and $N = \beta \text{diag}(0, 1)$ predicts stability for $|\delta| < 0.125$. The Popov bound predicts stability for $|\delta| < 0.125$ with increased performance. Finally, the shifted Popov bound predicts stability for $|\delta| < 0.425$ and coincides with the performance prediction of the LMI bound.

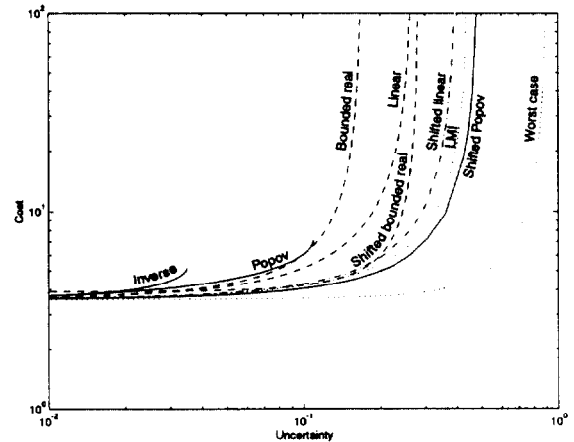


Figure 2: Cost bounds for Example 2.

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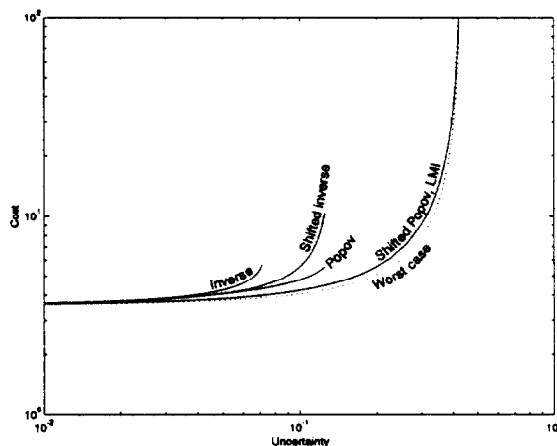


Figure 3: Cost bounds for Example 3.

10. Appendix

Proof of Theorem 1. Note that for all $\Delta A \in \mathcal{U}$, (5) is equivalent to

$$0 = (A + \Delta A)^T P (A + \Delta A) - P + \Omega(P) - (\Delta A^T P A + A^T P \Delta A + \Delta A^T P \Delta A) + R. \quad (42)$$

Adding and subtracting $(A + \Delta A)^T P_0(\Delta A)(A + \Delta A) - P_0(\Delta A)$ to (42) yields

$$0 = (A + \Delta A)^T (P + P_0(\Delta A))(A + \Delta A) - (P + P_0(\Delta A)) + \Omega(P) - [(A + \Delta A)^T P_0(\Delta A)(A + \Delta A) - P_0(\Delta A)] - (\Delta A^T P A + A^T P \Delta A + \Delta A^T P \Delta A) + R. \quad (43)$$

Hence, by assumption, (43) has a solution $P \in \mathcal{N}$ for all $\Delta A \in \mathbf{R}^{n \times n}$. If ΔA is restricted to the set \mathcal{U} , then, by (7), $\Omega(P) - [(A + \Delta A)^T P_0(\Delta A)(A + \Delta A) - P_0(\Delta A)] - (\Delta A^T P A + A^T P \Delta A + \Delta A^T P \Delta A)$ is nonnegative definite. Thus if $(A + \Delta A, E)$ is detectable for all $\Delta A \in \mathcal{U}$, then Theorem 3.6 of [25] implies $(A + \Delta A, [R + \hat{\Omega}(P, \Delta A) - (\Delta A^T P A + A^T P \Delta A + \Delta A^T P \Delta A)]^{1/2})$ is detectable for all $\Delta A \in \mathcal{U}$, where

$$\bar{\Omega}(P, \Delta A) \triangleq \Omega(P) - [(A + \Delta A)^T P_0(\Delta A)(A + \Delta A) - P_0(\Delta A)].$$

It now follows from (43) and Lemma 12.2' of [25] that $A + \Delta A$ is asymptotically stable for all $\Delta A \in \mathcal{U}$. Conversely, if $A + \Delta A$ is asymptotically stable for all $\Delta A \in \mathcal{U}$, then it follows immediately that $(A + \Delta A, E)$ is detectable for all $\Delta A \in \mathcal{U}$. Now, subtracting (4) from (43) yields

$$\begin{aligned} 0 &= (A + \Delta A)^T(P + P_0(\Delta A) - P_{\Delta A})(A + \Delta A) \\ &\quad - (P + P_0(\Delta A) - P_{\Delta A}) + \Omega(P) \\ &\quad - [(A + \Delta A)^T P_0(\Delta A)(A + \Delta A) - P_0(\Delta A)] \\ &\quad - (\Delta A^T P A + A^T P \Delta A + \Delta A^T P \Delta A), \quad \Delta A \in \mathcal{U} \quad (44) \end{aligned}$$

or, equivalently, since $A + \Delta A$ is asymptotically stable for all $\Delta A \in \mathcal{U}$

$$\begin{aligned} 0 &\leq \sum_{k=0}^{\infty} (A + \Delta A)^{kT} [\bar{\Omega}(P, \Delta A) \\ &\quad - (\Delta A^T P A + A^T P \Delta A + \Delta A^T P \Delta A)] (A + \Delta A)^k \\ &= P + P_0(\Delta A) - P_{\Delta A}, \quad \Delta A \in \mathcal{U} \end{aligned}$$

which implies (8). The performance bounds (9) and (11) are now an immediate consequence of (3), (8), and (10). \square

Proof of Proposition 1. Let $\Delta A \in \mathcal{U}$, and let

$$P = \hat{P} - (\bar{P}_0 - \tilde{P}_0). \quad (45)$$

Then since $P \in \mathcal{N}$, it follows that $\hat{P} \in \tilde{\mathcal{N}}$. Substituting (45) into (7), yields

$$\begin{aligned} 0 &\leq -\Delta A^T P A - A^T P \Delta A - \Delta A^T P \Delta A + \Omega(P) \\ &\quad - [(A + \Delta A)^T P_0(\Delta A)(A + \Delta A) - P_0(\Delta A)] \\ &= -\Delta A^T (\hat{P} - \bar{P}_0 + \tilde{P}_0) A - A^T (\hat{P} - \bar{P}_0 + \tilde{P}_0) \Delta A \\ &\quad - \Delta A^T (\hat{P} - \bar{P}_0 + \tilde{P}_0) \Delta A + \Omega(\hat{P} - \bar{P}_0 + \tilde{P}_0) \\ &\quad - [(A + \Delta A)^T P_0(\Delta A)(A + \Delta A) - P_0(\Delta A)] \\ &= -(\Delta A^T \hat{P} A + A^T \hat{P} \Delta A + \Delta A^T \hat{P} \Delta A) + \Omega(\hat{P} - \bar{P}_0 + \tilde{P}_0) \\ &\quad - [(A + \Delta A)^T P_0(\Delta A)(A + \Delta A) - P_0(\Delta A)] \\ &\quad + \Delta A^T (\bar{P}_0 - \tilde{P}_0) A + A^T (\bar{P}_0 - \tilde{P}_0) \Delta A + \Delta A^T (\bar{P}_0 - \tilde{P}_0) \Delta A \\ &= -(\Delta A^T \hat{P} A + A^T \hat{P} \Delta A + \Delta A^T \hat{P} \Delta A) + \hat{\Omega}(\hat{P}) \\ &\quad - [(A + \Delta A)^T P_0(\Delta A)(A + \Delta A) - P_0(\Delta A)] \\ &\quad + (A + \Delta A)^T (\bar{P}_0 - \tilde{P}_0)(A + \Delta A) - (\bar{P}_0 - \tilde{P}_0) \\ &= -(\Delta A^T \hat{P} A + A^T \hat{P} \Delta A + \Delta A^T \hat{P} \Delta A) + \hat{\Omega}(\hat{P}) \\ &\quad - [(A + \Delta A)^T (P_0(\Delta A) - \bar{P}_0 + \tilde{P}_0)(A + \Delta A) \\ &\quad - (P_0(\Delta A) - \bar{P}_0 + \tilde{P}_0)] \\ &= -(\Delta A^T \hat{P} A + A^T \hat{P} \Delta A + \Delta A^T \hat{P} \Delta A) + \hat{\Omega}(\hat{P}) \\ &\quad - [(A + \Delta A)^T \tilde{P}_0(\Delta A)(A + \Delta A) - \tilde{P}_0(\Delta A)] \end{aligned}$$

which completes the proof. \square

Proof of Proposition 4. Let $X = (T - B_0^T P B_0)$. Then

$$\begin{aligned} 0 &\leq \{X^{-1}(B_0^T P A - H) - X F C_0\}^T \{X^{-1}(B_0^T P A - H) - X F C_0\} \\ &\quad + Z - (H^T F C_0 + C_0^T F^T H + C_0^T F^T T F C_0) \\ &= (A^T P B_0 - H^T) X^{-2} (B_0^T P A - H) + C_0^T F^T X^2 F C_0 \\ &\quad - C_0^T F^T (B_0^T P A - H) + Z - (A^T P B_0 - H^T) F C_0 \\ &\quad - (H^T F C_0 + C_0^T F^T H + C_0^T F^T T F C_0) \\ &= (A^T P B_0 - H^T) (T - B_0^T P B_0)^{-1} (B_0^T P A - H) \\ &\quad + C_0^T F^T (T - B_0^T P B_0) F C_0 - C_0^T F^T B_0^T P A + Z \\ &\quad - A^T P B_0 F C_0 - C_0^T F^T T F C_0 \end{aligned}$$

$$\begin{aligned} &= (A^T P B_0 - H^T) (T - B_0^T P B_0)^{-1} (B_0^T P A - H) + Z \\ &\quad - C_0^T F^T B_0^T P A - A^T P B_0 F C_0 - C_0^T F^T B_0^T P B_0 F C_0 \\ &= (A^T P B_0 - H^T) (T - B_0^T P B_0)^{-1} (B_0^T P A - H) + Z \\ &\quad - \Delta A^T P A - A^T P \Delta A - \Delta A^T P \Delta A \\ &= \Omega(P) - \Delta A^T P A - A^T P \Delta A - \Delta A^T P \Delta A. \end{aligned}$$

which completes the proof. \square

Proof of Proposition 7. Define $X = (1 - \alpha)(P - N) \geq 0$, and $A_0 = \sum_{i=1}^r A_i$. Then

$$\begin{aligned} 0 &\leq \alpha^{-1} \left(\sum_{i=1}^r \delta_i A_i^T - A^T (P - N) \alpha X^{-1} \right) \\ &\quad \times X \left(\sum_{i=1}^r \delta_i A_i - \alpha X^{-1} (P - N) A \right) \\ &= \alpha A^T (P - N) X^{-1} (P - N) A + \alpha^{-1} \left(\sum_{i=1}^r \delta_i A_i^T \right) X \left(\sum_{i=1}^r \delta_i A_i \right) \\ &\quad - \Delta A^T (P - N) A - A^T (P - N) \Delta A \\ &\leq \frac{\alpha}{1 - \alpha} A^T (P - N) A + \gamma^2 \alpha^{-1} A_0^T (P - N) A_0 \\ &\quad + \gamma^2 A_0^T N A_0 + \gamma \sum_{i=1}^r |A_i^T N A + A^T N A_i| \\ &\quad - \Delta A^T P \Delta A - \Delta A^T P A - A^T P \Delta A \\ &= \Omega(P) - \Delta A^T P A - A^T P \Delta A - \Delta A^T P \Delta A. \end{aligned}$$

which completes the proof. \square

Proof of Proposition 11

$$\begin{aligned} 0 &\leq \left[H C_0 + (N C_0 + H^{-1} B_0^T C_0^T H \mu C_0) (A_{\text{Pop}} - I) \right. \\ &\quad \left. + H^{-1} B_0^T P A_{\text{Pop}} - X B_0^T \right] - R_0 H (F - M_L) C_0 \Big)^T R_0^{-1} \\ &\quad \times \left[H C_0 + (N C_0 + H^{-1} B_0^T C_0^T H \mu C_0) (A_{\text{Pop}} - I) \right. \\ &\quad \left. + H^{-1} B_0^T P A_{\text{Pop}} - X B_0^T \right] - R_0 H (F - M_L) C_0 \Big) \\ &\quad + \{(A_{\text{Pop}} - I) + \Delta A - B_0 M_L C_0\}^T (\bar{P}_0 - P_0(F)) \\ &\quad \times \{(A_{\text{Pop}} - I) + \Delta A - B_0 M_L C_0\} \\ &\quad + 2 C_0^T H \left((F - M_L) - (F - M_L) M^{-1} (F - M_L) \right) H C_0 \\ &\quad - \left[B_0 X^T (F - M_L) H C_0 + C_0^T H (F - M_L) X B_0^T \right] \\ &\quad - (A_{\text{Pop}} - I)^T C_0^T H \mu C_0 (A_{\text{Pop}} - I) + Z \\ &= \left[H C_0 + (N C_0 + H^{-1} B_0^T C_0^T H \mu C_0) (A_{\text{Pop}} - I) \right. \\ &\quad \left. + H^{-1} B_0^T P A_{\text{Pop}} - X B_0^T \right]^T R_0^{-1} \left[H C_0 - X B_0^T \right. \\ &\quad \left. + (N C_0 + H^{-1} B_0^T C_0^T H \mu C_0) (A_{\text{Pop}} - I) + H^{-1} B_0^T P A_{\text{Pop}} \right] \\ &\quad - P_0(F) (\Delta A - B_0 M_L C_0) - (\Delta A^T - C_0^T M_L B_0^T) P_0(F) \\ &\quad - (\Delta A^T - C_0^T M_L B_0^T) P (\Delta A - B_0 M_L C_0) \\ &\quad - \left[(A_{\text{Pop}}^T - I) P_0(F) + A_{\text{Pop}}^T P (\Delta A - B_0 M_L C_0) \right] \\ &\quad - \left[P_0(F) (A_{\text{Pop}} - I) + (\Delta A^T - C_0^T M_L B_0^T) P A_{\text{Pop}} \right] \\ &\quad - [A - I + \Delta A]^T P_0(F) [A - I + \Delta A] + Z \\ &= \left[H C_0 - X B_0^T + (N C_0 + H^{-1} B_0^T C_0^T H \mu C_0) (A_{\text{Pop}} - I) \right. \\ &\quad \left. + H^{-1} B_0^T P A_{\text{Pop}} \right]^T R_0^{-1} \left[H C_0 - X B_0^T \right. \\ &\quad \left. + (N C_0 + H^{-1} B_0^T C_0^T H \mu C_0) (A_{\text{Pop}} - I) + H^{-1} B_0^T P A_{\text{Pop}} \right] \\ &\quad + A^T P B_0 M_L C_0 + C_0^T M_L B_0^T P A + C_0^T M_L B_0^T P B_0 M_L C_0 \\ &\quad + Z - (\Delta A^T P \Delta A + A^T P \Delta A + \Delta A^T P A) \\ &\quad - (A + \Delta A)^T P_0(F) (A + \Delta A) + P_0(F) \\ &= \Omega_0(P) - (\Delta A^T P A + A^T P \Delta A + \Delta A^T P \Delta A) \\ &\quad - [(A + \Delta A)^T P_0(F) (A + \Delta A) - P_0(F)] \end{aligned}$$

which completes the proof. \square