

SOME NEW RESULTS ON STABILITY AND STABILITY ROBUSTNESS OF FORNASINI-MARCHESINI STATE-SPACE 2-D DIGITAL FILTERS

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Abstract- This paper describes a Lyapunov approach to stability and stability robustness analysis for 2-D digital filters in the Fornasini-Marchesini local state-space setting. A class of new constant 2-D Lyapunov equations, that generalize the 2-D Lyapunov equation proposed recently by T. Hinamoto, is described. It is shown that the generalized 2-D Lyapunov equation can be used to derive a new stability condition that narrows the gap between the “necessity” and “sufficiency” that occurs in Hinamoto’s stability theorem. The generalized 2-D Lyapunov equations are then utilized to derive two lower bounds of unstructured, stable perturbations of a given stable filter. Numerical techniques for solving the proposed 2-D Lyapunov equations are also presented.

I. INTRODUCTION

Various Lyapunov equations have been used as effective tool for stability analysis of both one-dimensional (1-D) and two-dimensional (2-D) systems, see e.g. [1]–[8] and the references cited there. Unlike its 1-D counterpart, however, the stability theorems based on the constant 2-D Lyapunov equation investigated in [4][7] for the Roesser model and the constant 2-D Lyapunov equation proposed and studied in [8] for the Fornasini-Marchesini model provide only *sufficient* conditions for stability of the state-space digital filters considered. In particular, as can be seen from [8] and the analysis presented in Sec. II of this paper, there are many 2-D filters that are stable but do not satisfy the Lyapunov equation proposed in [8]. One of the chief objectives of this paper is to propose a generalized constant 2-D Lyapunov equation and to apply it in the stability analysis of the 2-D digital filters in the Fornasini-Marchesini local state-space setting. Again the stability theorem obtained here gives a sufficient con-

dition, but it includes Hinamoto’s stability theorem [8] as a special case and narrows the gap between “sufficiency” and “necessity” for a state-space digital filter to be stable, which occurs in Hinamoto’s Lyapunov theorem. In addition, we make use of the generalized 2-D Lyapunov equation to analyze the robust stability of a F-M 2-D model, leading to two lower bounds of the unstructured, stable perturbations.

II. PRELIMINARIES

The 2-D digital filters considered here are those modeled in Fornasini and Marchesini’s local state space by

$$\begin{aligned} \mathbf{x}(i+1, j+1) &= \mathbf{A}_1 \mathbf{x}(i, j+1) + \mathbf{A}_2 \mathbf{x}(i+1, j) \\ &\quad + \mathbf{b}_1 u(i, j+1) + \mathbf{b}_2 u(i+1, j) \quad (1a) \\ y(i, j) &= \mathbf{c} \mathbf{x}(i, j) + d u(i, j) \quad (1b) \end{aligned}$$

where $\mathbf{A}_1, \mathbf{A}_2 \in R^{n \times n}$. It is known [9] that (1) is asymptotically stable if and only if

$$\begin{aligned} p(z_1, z_2) &\equiv \det(\mathbf{I} - z_1 \mathbf{A}_1 - z_2 \mathbf{A}_2) \neq 0 \\ \text{for } (z_1, z_2) &\in \bar{U} \end{aligned} \quad (2)$$

where $\bar{U} = \{(z_1, z_2) : |z_1| \leq 1, |z_2| \leq 1\}$. In [8] Hinamoto proved the following theorem:

Theorem 1 [8]: (1) is asymptotically stable if there exists a positive definite symmetric $\mathbf{P} \in R^{n \times n}$ such that

$$\mathbf{Q} = \begin{bmatrix} \alpha \mathbf{P} & \mathbf{0} \\ \mathbf{0} & \beta \mathbf{P} \end{bmatrix} - \mathbf{A}^T \mathbf{P} \mathbf{A} \quad (3)$$

is positive definite where $\mathbf{A} = [\mathbf{A}_1 \quad \mathbf{A}_2]$, and

$$\alpha, \beta > 0, \quad \alpha + \beta = 1 \quad (4)$$

A graphic illustration of Theorem 1 was also given in [8] as follows. If filter (1) satisfies (3) and (4), then all zeros of characteristic polynomial $p(z_1, z_2)$ correspond to

locations outside the ellipse in the first quadrant shown in Fig. 1, where the ellipse is described by

$$\frac{|z_1|^2}{1/\alpha} + \frac{|z_2|^2}{1/\beta} = 1 \quad (5)$$

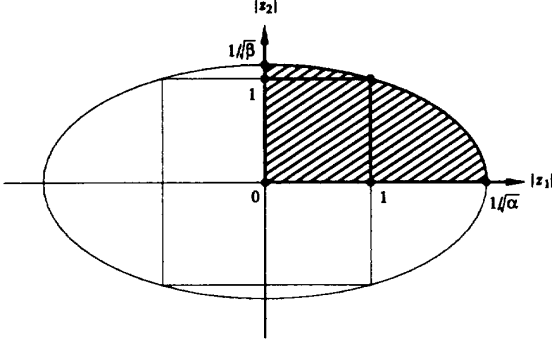


Figure 1: A graphic illustration of the conditions in Theorem 1 [8].

From this illustration it follows that stable filters with poles (i.e. zeros of $p(z_1, z_2)$) corresponding to the locations in *both* shadowed areas S_1 and S_2 shown in Fig. 2, for example, do not satisfy (3) and (4). This is to say, no α and β can be found to meet (4) such that zeros of $p(z_1, z_2)$ in both areas S_1 and S_2 are left outside the corresponding ellipse. Thus we see that there is a large number of filters that are stable but their stability cannot be confirmed using Theorem 1.

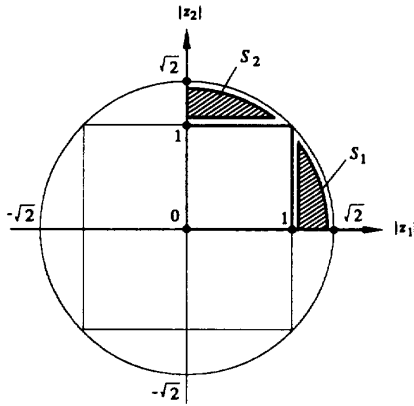


Figure 2: A graphic illustration of the sufficiency of Theorem 1.

Note that (3) can be written as

$$\mathbf{Q} = \begin{bmatrix} \mathbf{P}^{T/2}(\alpha\mathbf{I})\mathbf{P}^{1/2} & 0 \\ 0 & \mathbf{P}^{T/2}(\beta\mathbf{I})\mathbf{P}^{1/2} \end{bmatrix} - \mathbf{A}^T \mathbf{P}^{T/2} \mathbf{I} \mathbf{P}^{1/2} \mathbf{A} \quad (6)$$

In the rest of the paper we shall consider the Lyapunov equation

$$\mathbf{Q} = \begin{bmatrix} \mathbf{P}^{T/2} \mathbf{W}_1 \mathbf{P}^{1/2} & 0 \\ 0 & \mathbf{P}^{T/2} \mathbf{W}_2 \mathbf{P}^{1/2} \end{bmatrix} - \mathbf{A}^T \mathbf{P}^{T/2} \mathbf{R} \mathbf{P}^{1/2} \mathbf{A} \quad (7)$$

where matrices $\mathbf{P}$, $\mathbf{W}_1$, $\mathbf{W}_2$ and $\mathbf{R}$ are all positive definite. Clearly (7) becomes (3) if $\mathbf{W}_1 = \alpha\mathbf{I}$, $\mathbf{W}_2 = \beta\mathbf{I}$ and $\mathbf{R} = \mathbf{I}$. For this reason we shall call (7) the *generalized 2-D Lyapunov equation* associated with state-space model (1).

III. A GENERALIZED 2-D LYAPUNOV THEORY

The main theorem, stated and proved below, describes the use of the generalized Lyapunov equation (7) to confirm stability of a 2-D state-space digital filter.

Theorem 2: (1) is asymptotically stable if there are positive definite matrices $\mathbf{P}$, $\mathbf{W}_1$, $\mathbf{W}_2$, and $\mathbf{R}$, such that matrix $\mathbf{Q}$ given by (7) is positive definite and that

$$\mathbf{R} - \mathbf{W}_1 - \mathbf{W}_2 \geq 0 \quad (8)$$

Proof: We prove the theorem by contradiction. Suppose the conditions of the theorem are satisfied but (1) is unstable. Then there is $(z_1, z_2) \in \bar{U}$ such that

$$\det(\mathbf{I} - z_1 \mathbf{A}_1 - z_2 \mathbf{A}_2) = 0 \quad (9)$$

Hence there exists $\mathbf{v} \neq 0$ such that

$$\mathbf{v} = \mathbf{A} \begin{bmatrix} z_1 \mathbf{I} \\ z_2 \mathbf{I} \end{bmatrix} \mathbf{v} \quad (10)$$

where $\mathbf{A} = [\mathbf{A}_1 \quad \mathbf{A}_2]$. We now use (7) and (10) to compute

$$\begin{aligned} & \mathbf{v}^* \mathbf{P}^{T/2} \mathbf{R} \mathbf{P}^{1/2} \mathbf{v} \\ &= \mathbf{v}^* \begin{bmatrix} \bar{z}_1 \mathbf{I} & \bar{z}_2 \mathbf{I} \end{bmatrix} \begin{bmatrix} \mathbf{P}^{T/2} \mathbf{W}_1 \mathbf{P}^{1/2} & 0 \\ 0 & \mathbf{P}^{T/2} \mathbf{W}_2 \mathbf{P}^{1/2} \end{bmatrix} \\ & \quad \cdot \begin{bmatrix} z_1 \mathbf{I} \\ z_2 \mathbf{I} \end{bmatrix} \mathbf{v} - \mathbf{v}_z^* \mathbf{Q} \mathbf{v}_z \end{aligned}$$

where $\mathbf{v}^*$ denotes the complex conjugate transposition of $\mathbf{v}$, $\bar{z}_i$ is the complex conjugate of z_i , and $\mathbf{v}_z^* = \mathbf{v}^* \begin{bmatrix} \bar{z}_1 \mathbf{I} & \bar{z}_2 \mathbf{I} \end{bmatrix}$. It follows that

$$\mathbf{v}^* \mathbf{P}^{T/2} (\mathbf{R} - |z_1|^2 \mathbf{W}_1 - |z_2|^2 \mathbf{W}_2) \mathbf{P}^{1/2} \mathbf{v} = -\mathbf{v}_z^* \mathbf{Q} \mathbf{v}_z \quad (11)$$

By (9) we note that $(z_1, z_2) \neq (0, 0)$, hence $\mathbf{v}_z \neq 0$. So $\mathbf{Q} > 0$ means that the right-hand side of (11) is negative. On the other hand, $|z_1| \leq 1$, $|z_2| \leq 1$ and the positive semi-definiteness of $\mathbf{R} - \mathbf{W}_1 - \mathbf{W}_2$ imply that $\mathbf{R} - |z_1|^2 \mathbf{W}_1 - |z_2|^2 \mathbf{W}_2 \geq 0$, therefore the left-hand side of (11) is nonnegative, leading to a contradiction. This completes the proof. $\square$

Several consequences can be drawn from Theorem 2 immediately.

Corollary 2.1: (1) is asymptotically stable if there is a matrix $\mathbf{P} > 0$ such that

$$\mathbf{Q} = \begin{bmatrix} \mathbf{P}^{T/2} \mathbf{W}_1 \mathbf{P}^{1/2} & 0 \\ 0 & \mathbf{P}^{T/2} \mathbf{W}_2 \mathbf{P}^{1/2} \end{bmatrix} - \mathbf{A}^T \mathbf{P} \mathbf{A} \quad (12)$$

is positive definite, where $\mathbf{W}_1 = \mathbf{U}^T \Sigma_1 \mathbf{U}$, $\mathbf{W}_2 = \mathbf{U}^T \Sigma_2 \mathbf{U}$ with $\mathbf{U}$ orthogonal, $\Sigma_1 = \text{diag}\{\sigma_{11}, \dots, \sigma_{1n}\}$, $\Sigma_2 = \text{diag}\{\sigma_{21}, \dots, \sigma_{2n}\}$,

$$\sigma_{1k} > 0, \sigma_{2k} > 0 \text{ and } \sigma_{1k} + \sigma_{2k} = 1 \quad 1 \leq k \leq n \quad (13)$$

Corollary 2.2: (1) is asymptotically stable if there is a matrix $\mathbf{P} > 0$ such that

$$\mathbf{Q} = \begin{bmatrix} \mathbf{P}^{T/2} \mathbf{W}_1 \mathbf{P}^{1/2} & 0 \\ 0 & \mathbf{P}^{T/2} \mathbf{W}_2 \mathbf{P}^{1/2} \end{bmatrix} - \mathbf{A}^T \mathbf{P} \mathbf{A} \quad (14)$$

is positive definite, where $\mathbf{W}_i = \text{diag}\{\sigma_{i1}, \dots, \sigma_{in}\}$ ($i = 1, 2$) satisfy (13).

Corollary 2.3 (Hinamoto): (1) is asymptotically stable if there is a matrix $\mathbf{P} > 0$ such that

$$\mathbf{Q} = \begin{bmatrix} \alpha \mathbf{P} & 0 \\ 0 & \beta \mathbf{P} \end{bmatrix} - \mathbf{A}^T \mathbf{P} \mathbf{A} \quad (15)$$

is positive definite, where

$$\alpha > 0, \beta > 0 \text{ and } \alpha + \beta = 1 \quad (16)$$

Note that in all three corollaries matrix $\mathbf{R}$ is assigned to be the identity matrix. As will be seen from the next theorem, this assignment does not lose the generality of Theorem 2.

Theorem 3: The sufficient conditions stated in Theorem 2 for (1) to be asymptotically stable and the conditions stated below are equivalent: There exist matrices $\mathbf{P} > 0$, $\mathbf{W}_1 > 0$, and $\mathbf{W}_2 > 0$ such that

$$\mathbf{Q} = \begin{bmatrix} \mathbf{P}^{T/2} \mathbf{W}_1 \mathbf{P}^{1/2} & 0 \\ 0 & \mathbf{P}^{T/2} \mathbf{W}_2 \mathbf{P}^{1/2} \end{bmatrix} - \mathbf{A}^T \mathbf{P} \mathbf{A} \quad (17)$$

is positive definite and

$$\mathbf{I} - \mathbf{W}_1 - \mathbf{W}_2 \geq 0 \quad (18)$$

The proof of this theorem can be found in [10].

IV. TWO LOWER BOUNDS OF ν_1

Consider a stable system (1a) with $\mathbf{b}_1 = 0$, $\mathbf{b}_2 = 0$ that satisfies (17) and (18). Let $\mathbf{A} = [\mathbf{A}_1 \ \mathbf{A}_2]$ and denote by S_u the set of all perturbations $\Delta \mathbf{A} = [\Delta \mathbf{A}_1 \ \Delta \mathbf{A}_2] \in \mathbb{C}^{n \times 2n}$ with $\mathbf{A} + \Delta \mathbf{A}$ unstable, i.e.,

$$S_u = \{\Delta \mathbf{A} : [\Delta \mathbf{A}_1 \ \Delta \mathbf{A}_2] \in \mathbb{C}^{n \times 2n}, \mathbf{A} + \Delta \mathbf{A} \text{ unstable}\}$$

A measure for robust stability of system (1) is defined as

$$\nu_1 = \inf_{\Delta \mathbf{A} \in S_u} \|\Delta \mathbf{A}\|$$

Let the perturbed system be given by

$$\mathbf{x}(i+1, j+1) = (\mathbf{A}_1 + \Delta \mathbf{A}_1) \mathbf{x}(i, j+1) + (\mathbf{A}_2 + \Delta \mathbf{A}_2) \mathbf{x}(i+1, j) \quad (19)$$

The matrices $\mathbf{P}$, $\mathbf{W}_1$ and $\mathbf{W}_2$ are employed to construct the Lyapunov function

$$\phi(i, j) = \phi_1(i, j) + \phi_2(i, j) \quad (20)$$

where

$$\phi_1(i, j) = \mathbf{x}^T(i, j+1) \mathbf{P}^{T/2} \mathbf{W}_1 \mathbf{P}^{1/2} \mathbf{x}(i, j+1) \quad (21)$$

$$\phi_2(i, j) = \mathbf{x}^T(i+1, j) \mathbf{P}^{T/2} \mathbf{W}_2 \mathbf{P}^{1/2} \mathbf{x}(i+1, j) \quad (22)$$

Define

$$\phi_{11}(i, j) = \phi_1(i+1, j) + \phi_2(i, j+1) \quad (23)$$

By (18),

$$\begin{aligned} \phi_{11}(i, j) &= \mathbf{x}^T(i+1, j+1) \\ &\quad \cdot \mathbf{P}^{T/2} (\mathbf{W}_1 + \mathbf{W}_2) \mathbf{P}^{1/2} \mathbf{x}(i+1, j+1) \\ &\leq \mathbf{x}^T(i+1, j+1) \mathbf{P} \mathbf{x}(i+1, j+1) \end{aligned} \quad (24)$$

Now using (17) and (19)-(24) we compute

$$\begin{aligned} \Delta \phi(i, j) &= \phi_{11}(i, j) - \phi(i, j) \\ &\leq -[\underline{\sigma}(\mathbf{Q}) - 2\bar{\sigma}^{1/2}(\tilde{\mathbf{P}} - \mathbf{Q})\bar{\sigma}^{1/2}(\mathbf{P})\|\Delta \mathbf{A}\| \\ &\quad - \bar{\sigma}(\mathbf{P})\|\Delta \mathbf{A}\|^2] \|\tilde{\mathbf{x}}(i, j)\|^2 \end{aligned}$$

where

$$\begin{aligned} \tilde{\mathbf{x}}(i, j) &= \begin{bmatrix} \mathbf{x}(i, j+1) \\ \mathbf{x}(i+1, j) \end{bmatrix} \\ \text{and } \tilde{\mathbf{P}} &= \begin{bmatrix} \mathbf{P}^{T/2} \mathbf{W}_1 \mathbf{P}^{T/2} & 0 \\ 0 & \mathbf{P}^{T/2} \mathbf{W}_2 \mathbf{P}^{1/2} \end{bmatrix} \end{aligned} \quad (25)$$

Hence if

$$\|\Delta \mathbf{A}\| < \frac{[\bar{\sigma}(\tilde{\mathbf{P}} - \mathbf{Q}) + \underline{\sigma}(\mathbf{Q})]^{1/2} - \bar{\sigma}^{1/2}(\tilde{\mathbf{P}} - \mathbf{Q})}{\bar{\sigma}^{1/2}(\mathbf{P})} \quad (26)$$

then $\Delta\phi(i, j) < 0$ for all i, j , and a typical argument (see e.g., [4]) applies to show the stability of the perturbed system (19).

An alternative Lyapunov function may be defined as

$$\psi(i, j) = [\phi_1(i, j) + \phi_2(i, j)]^{1/2} \quad (27)$$

Similarly we define

$$\psi_{11}(i, j) = [\phi_1(i+1, j) + \phi_2(i, j+1)]^{1/2} \quad (28)$$

and use (18) to write

$$\psi_{11}(i, j) \leq [\mathbf{x}^T(i+1, j+1)\mathbf{P}\mathbf{x}(i+1, j+1)]^{1/2} \quad (29)$$

Now (17), (19), (27) and (29) imply that

$$\begin{aligned} \Delta\psi(i, j) &= \psi_{11}(i, j) - \psi(i, j) \\ &\leq -\left[\frac{\underline{\sigma}(\mathbf{Q})}{\bar{\sigma}^{1/2}(\tilde{\mathbf{P}} - \mathbf{Q}) + \bar{\sigma}^{1/2}(\tilde{\mathbf{P}})} - \bar{\sigma}^{1/2}(\mathbf{P})\|\Delta\mathbf{A}\| \right] \|\tilde{\mathbf{x}}(i, j)\| \end{aligned}$$

Thus if

$$\|\Delta\mathbf{A}\| < \frac{\underline{\sigma}(\mathbf{Q})}{\bar{\sigma}^{1/2}(\mathbf{P})[\bar{\sigma}^{1/2}(\tilde{\mathbf{P}} - \mathbf{Q}) + \bar{\sigma}^{1/2}(\tilde{\mathbf{P}})]} \quad (30)$$

then $\Delta\psi(i, j) < 0$ for all i, j , and therefore, the perturbed system (19) will remain stable.

Although our numerical experience indicates that the lower bounds of ν_i given in (26) and (30) usually stay quite close each other, the bound given in (26) is always better than that in (30) [11].

V. NUMERICAL SOLUTION TO THE GENERALIZED LYAPUNOV EQUATION

In this section we consider the problem of numerically solving Lyapunov equation (17), (18) in the following sense: Given $\mathbf{A} = [\mathbf{A}_1 \ \mathbf{A}_2]$, find $\mathbf{P} > 0$, $\mathbf{W}_1 > 0$, and $\mathbf{W}_2 > 0$ such that $\mathbf{Q}$ defined by (17) is positive definite and (18) holds. In what follows, we first present a result that relates the existence of such positive definite $\mathbf{P}$, $\mathbf{W}_1$, $\mathbf{W}_2$ to a norm minimization problem. Discussions then follow to consider numerical solution of this norm minimization problem for two special cases where both matrices $\mathbf{W}_1$, $\mathbf{W}_2$ are diagonal.

Theorem 4: There exist $\mathbf{P} > 0$, $\mathbf{W}_1 > 0$ and $\mathbf{W}_2 > 0$ such that (18) holds and $\mathbf{Q}$ defined in (17) is positive definite if and only if

$$\min_{\substack{\mathbf{T}, \mathbf{V}_1, \mathbf{V}_2 \text{ nonsingular} \\ \mathbf{I} - \mathbf{V}_1^{-T} \mathbf{r} \mathbf{V}_1^{-1} - \mathbf{V}_2^{-T} \mathbf{r} \mathbf{V}_2^{-1} \geq 0}} \left\| \tilde{\mathbf{A}} \begin{bmatrix} \mathbf{V}_1 & \mathbf{0} \\ \mathbf{0} & \mathbf{V}_2 \end{bmatrix} \right\| < 1 \quad (31)$$

where $\tilde{\mathbf{A}} = [\tilde{\mathbf{A}}_1 \ \tilde{\mathbf{A}}_2]$, $\tilde{\mathbf{A}}_1 = \mathbf{T}^{-1} \mathbf{A}_1 \mathbf{T}$, $\tilde{\mathbf{A}}_2 = \mathbf{T}^{-1} \mathbf{A}_2 \mathbf{T}$, and $\|\cdot\|$ denotes the induced 2-norm of the matrix involved.

Proof: see [10].

There is a variety of efficient methods that can be used to solve the optimization problem (31) [12][13]. In what follows we discuss two special cases where $\mathbf{V}_1$ and $\mathbf{V}_2$ assume diagonal forms, leading to much simplified unconstrained minimization problems.

Case A: In this case $\mathbf{V}_1$ and $\mathbf{V}_2$ assume the form of

$$\mathbf{V}_1 = \begin{bmatrix} 1/v_{11} & & \mathbf{0} \\ & \ddots & \\ \mathbf{0} & & 1/v_{1n} \end{bmatrix}, \quad \mathbf{V}_2 = \begin{bmatrix} 1/v_{21} & & \mathbf{0} \\ & \ddots & \\ \mathbf{0} & & 1/v_{2n} \end{bmatrix}$$

with $v_{1i} > 0$ and $v_{2i} > 0$ ($1 \leq i \leq n$). To meet constraint (28), it is required that

$$v_{1i}^2 + v_{2i}^2 = 1 \quad 1 \leq i \leq n$$

which implies that

$$\begin{aligned} 0 &< v_{1i} < 1 \\ v_{2i} &= (1 - v_{1i}^2)^{1/2} \end{aligned} \quad (32)$$

The hyperbolic tangent transformation

$$v_{1i} = \frac{1 + \tanh \nu_i}{2}, \quad 1 \leq i \leq n$$

can be used to meet constraint (32) with $\nu_i \in (-\infty, \infty)$.

The minimization problem is now simplified as

$$\min_{\substack{\mathbf{T} \text{ nonsingular} \\ -\infty < \nu_1, \dots, \nu_n < \infty}} \left\| \tilde{\mathbf{A}} \begin{bmatrix} \mathbf{V}_1 & \mathbf{0} \\ \mathbf{0} & \mathbf{V}_2 \end{bmatrix} \right\| \quad (33)$$

where

$$\begin{aligned} \mathbf{V}_1 &= \begin{bmatrix} \frac{2}{1 + \tanh \nu_1} & & \mathbf{0} \\ & \ddots & \\ \mathbf{0} & & \frac{2}{1 + \tanh \nu_n} \end{bmatrix} \\ \mathbf{V}_2 &= \begin{bmatrix} \frac{2}{[4 - (1 + \tanh \nu_1)^2]^{1/2}} & & \mathbf{0} \\ & \ddots & \\ \mathbf{0} & & \frac{2}{[4 - (1 + \tanh \nu_n)^2]^{1/2}} \end{bmatrix} \end{aligned}$$

Although the state-variable transformation matrix $\mathbf{T}$ in problem (33) needs to be nonsingular, (33) can be considered as an *unconstrained* optimization problem due to the following two reasons: First, the measure of the set

of points in the parameter space that make $\mathbf{T}$ singular is zero from a classic measure-theoretic point of view [14] meaning that the chance to meet a point in the parameter space that corresponds to a singular matrix $\mathbf{T}$ in a search process during minimization is extremely rare; second, during the minimization, updating a parameter vector to a new one that would be closer to a singular $\mathbf{T}$ is always discouraged as a nearly singular $\mathbf{T}$ tends to increase the value of the norm in (33) due to the presence of $\mathbf{T}^{-1}$ in $\tilde{\mathbf{A}}$. In other words, in this norm minimization problem the use of a good optimization algorithm would very unlikely lead to a solution that corresponds to a nearly singular $\mathbf{T}$.

Case B: In this case $\mathbf{V}_1$ and $\mathbf{V}_2$ assume the form of

$$\mathbf{V}_1 = \frac{1}{v_1} \mathbf{I} \text{ and } \mathbf{V}_2 = \frac{1}{v_2} \mathbf{I} \quad (34)$$

with $v_i > 0$ ($i = 1, 2$). To satisfy the constraint in (31), it is required that

$$v_1^2 + v_2^2 = 1$$

Hence

$$0 < v_1 < 1 \text{ and } v_2 = (1 - v_1^2)^{1/2}$$

Similar to Case A, hyperbolic tangent transformations

$$v_1 = \frac{1 + \tanh \nu}{2}, \quad -\infty < \nu < \infty$$

and

$$v_2 = \frac{1}{2} [4 - (1 + \tanh \nu)^2]^{1/2}, \quad -\infty < \nu < \infty$$

can be used in (34) with $-\infty < \nu < \infty$ so that the minimization problem

$$\underset{\substack{\mathbf{T} \text{ nonsingular} \\ -\infty < \nu < \infty}}{\text{minimize}} \left\| \tilde{\mathbf{A}} \begin{bmatrix} \mathbf{V}_1 & 0 \\ 0 & \mathbf{V}_2 \end{bmatrix} \right\| \quad (35)$$

with $\mathbf{V}_i$ ($i = 1, 2$) specified above is essentially unconstrained. Finally, it is easy to see that as $\mathbf{W}_1 = \mathbf{V}_1^{-T} \mathbf{V}_1^{-1} = v_1^2 \mathbf{I}$ and $\mathbf{W}_2 = \mathbf{V}_2^{-T} \mathbf{V}_2^{-1} = v_2^2 \mathbf{I}$, (35) provides a feasible approach to numerically solving Hinamoto's Lyapunov equation (3), (4).

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