

# ESKİŞEHİR TEKNİK ÜNİVERSİTESİ BİLİM VE TEKNOLOJİ DERGİSİ B- TEORİK BİLİMLER

Eskişehir Technical University Journal of Science and Technology B- Theoretical Sciences

2022, 10(1), pp. 18-26, DOI:10.20290/estubtdb.1057006

## **RESEARCH ARTICLE**

## COMMON QUADRATIC LYAPUNOV FUNCTIONS FOR TWO STABLE MATRICES

# Şerife YILMAZ \* 🔟

## ABSTRACT

In this paper we consider the problem of existence and evaluation of common solutions to Lyapunov equations for switched systems consisting of two real or complex matrices. Conditions for existence and solution algorithm based on the gradient of a matrix function are given. Number of examples are provided.

Keywords: Lyapunov function, Common solution, Hurwitz stability

## **1. INTRODUCTION**

Let  $\mathbb{R}^{n \times n}$  ( $\mathbb{C}^{n \times n}$ ) be the set of  $n \times n$  real (complex) matrices. A matrix A is called Hurwitz stable if its all eigenvalues lie in the open left-half plane. A necessary and sufficient condition for the matrix  $A \in \mathbb{R}^{n \times n}$  ( $A \in \mathbb{C}^{n \times n}$ ) to be Hurwitz stable is that for any given positive definite matrix Q there exists a (unique) positive definite matrix P that satisfies the Lyapunov equation.

Consider the following linear switched system

$$\dot{x} = Ax, \qquad A \in \{A_1, A_2, \dots, A_N\}$$
 (1)

where  $A_i \in \mathbb{R}^{n \times n}$   $(A_i \in \mathbb{C}^{n \times n})$  (i = 1, 2, ..., N) are Hurwitz stable.

If there exists a P > 0 such that

$$A_i^T P + P A_i < 0, \quad (A_i^* P + P A_i < 0) \quad (i = 1, 2, ..., N)$$
 (2)

the matrix P is called a common solution to Lyapunov inequalities (2) and the function  $V(x) = x^T P x$ ( $V(x) = x^* P x$ ) is called common quadratic Lyapunov function of the switched system (1). A sufficient condition for the switched system (1) to be global uniformly asimptotically stable is that the inequalities (2) have a common solution P (see [1,2]).

The problem of existence and evaluation of common solution to system (2) have been studied in a lot of works (see [1-6] and references therein). Theoretical condition on the existence of a solution to this problem for two real  $2 \times 2$  dimensional matrices is given in [3]. In the general *n*-dimensional case, there is no theoretical solution in the literature except for the theoretical results that can be given without resorting to numerical methods for the existence of the solution (see [3]). In [4] for a pair of  $2 \times 2$  complex matrices a necessary and sufficient condition for the existence of common solution *P* is given. This result relates to a special class of  $4 \times 4$  real matrices.

<sup>\*</sup>Corresponding Author: <u>serifeyilmaz@mehmetakif.edu.tr</u> Received: 20.12.2021 Published: 25.02.2022

In this paper we give a necessary and sufficient condition for the existence of a common solution for two  $n \times n$  dimensional real matrix. In the second part, we propose a gradient algorithm for common solution of Lyapunov inequalities for two  $n \times n$  dimensional complex matrix. Here we establish a formula for the derivative of a matrix functional (Theorem 2).

#### 2. CONDITION FOR COMMON SOLUTION FOR TWO REAL MATRICES

Let A be Hurwitz stable real matrix. For a given positive definite matrix Q, we denote the unique P > 0 solution of the equation  $A^T P + PA = -Q$  by  $P_{A(Q)}$ .

Let two  $n \times n$  dimensional Hurwitz stable real matrices  $A_1$  and  $A_2$  be given. For  $Q_1 > 0$  and  $Q_2 > 0$  we define

$$P_1 = P_{A_1(Q_1)} > 0, \qquad P_2 = P_{A_2(Q_2)}$$

and

$$l_{ij} = \lambda_{\max}(A_i^T P_j + P_j A_i), \qquad (i, j = 1, 2)$$

where  $\lambda_{\max}(C)$  stands for the maximum eigenvalue of C. From stability of  $A_1$  and  $A_2$  it follows that

$$l_{11} < 0 \text{ and } l_{22} < 0. \tag{3}$$

Define 2  $\times$  2 dimensional matrix

$$L = \begin{pmatrix} l_{11} & l_{12} \\ l_{21} & l_{22} \end{pmatrix}.$$

If  $l_{21} < 0$  then  $A_2^T P_1 + P_1 A_2 < 0$  therefore  $P_1$  is a common solution for  $\{A_1, A_2\}$ . If  $l_{12} < 0$  then  $A_1^T P_2 + P_2 A_1 < 0$  and  $P_2$  is a common solution. Therefore we assume that

$$l_{12} \ge 0 \quad \text{and} \quad l_{21} \ge 0.$$
 (4)

We investigate a weighted sum  $P_1$  and  $P_2$  for a common solution.

**Lemma 1.** Let the pair  $(w_1, w_2)$  is given where

 $l_{11}w_1 + l_{12}w_2 < 0, \quad l_{21}w_1 + l_{22}w_2 < 0, \quad w_1 > 0, \quad w_2 > 0.$  (5) Then  $P_* = w_1P_1 + w_2P_2$  is a common solution for  $\{A_1, A_2\}$ .

**Proof.** Using convexity of the function  $P \mapsto \lambda_{\max}(A_i^T P + PA_i)$  (see [5, p.34], [6]), we obtain

$$\lambda_{\max}(A_i^T P_* + P_* A_i) = \lambda_{\max}(w_1(A_i^T P_1 + P_1 A_i) + w_2(A_i^T P_2 + P_2 A_i))$$
  
$$\leq w_1 \lambda_{\max}(A_i^T P_1 + P_1 A_i) + w_2(A_i^T P_2 + P_2 A_i)$$
  
$$= w_1 l_{i1} + w_2 l_{i2} < 0 \quad (i = 1, 2).$$

Therefore the matrix  $P_* = w_1 P_1 + w_2 P_2$  is a common solution.

By (5) we consider for a positive solution of the system

$$\begin{cases} l_{11}x + l_{12}y < 0\\ l_{21}x + l_{22}y < 0 \end{cases}$$
(6)

If  $l_{12} = 0$  then any pair (x, y), where x > 0 is arbitrary and  $y > -\frac{l_{21}}{l_{22}}x$  is a feasible pair. If  $l_{21} = 0$  then any pair (x, y), where y > 0 is arbitrary and  $x > -\frac{l_{12}}{l_{11}}y$  is a feasible pair. Therefore we assume that

$$l_{12} > 0 \text{ and } l_{21} > 0.$$
 (7)

**Theorem 1.** Assume that (7) is satisfied. Then (6) has positive solution (x, y) if and only if

$$l_{11}l_{22} > l_{12}l_{21}.$$
 (8)

**Proof.** From (6) it follows that

$$y < -\frac{l_{11}}{l_{12}}x, \quad y > -\frac{l_{21}}{l_{22}}x, \quad \text{or} \quad y < k_1x, \quad y > k_2x,$$

where  $k_1 = -\frac{l_{11}}{l_{12}} > 0$ ,  $k_2 = -\frac{l_{21}}{l_{22}} > 0$ . Then (6) has a positive solution pair (x, y) if and only if  $k_1 > k_2$  (see Fig. 1). The inequality  $k_1 > k_2$  is equivalent to (8).

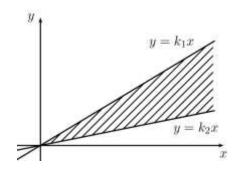


Figure 1.

If (8) is satisfied the weighted coefficients

$$w_1^0 = \frac{l_{12} - l_{22}}{l_{11}l_{22} - l_{21}l_{12}}, \quad w_2^0 = \frac{l_{21} - l_{11}}{l_{11}l_{22} - l_{21}l_{12}}$$

satisfy the inequalities

$$w_1^0 > 0$$
,  $w_2^0 > 0$ ,  $l_{11}w_1^0 + l_{12}w_2^0 = -1 < 0$ ,  $l_{21}w_1^0 + l_{22}w_2^0 = -1 < 0$ 

and the matrix  $P_* = w_1^0 P_1 + w_2^0 P_2$  is a common solution.

Example 1. Consider the following Hurwitz stable matrices

$$A_{1} = \begin{bmatrix} -5 & 2 & -1 & 2 \\ 2 & -4 & 1 & 2 \\ 3 & -2 & -2 & -3 \\ 5 & -2 & 1 & 7 \end{bmatrix}, A_{2} = \begin{bmatrix} -1 & -2 & -3 & 2 \\ -2 & -7 & -3 & 4 \\ 4 & -2 & -5 & 1 \\ -3 & 0 & -1 & -1 \end{bmatrix}$$

and positive definite matrices

$$Q_1 = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}, Q_2 = \begin{bmatrix} 12 & -1 & 2 & -3 \\ -1 & 11 & 1 & 0 \\ 2 & 0 & 11 & 0 \\ -3 & 0 & 2 & 11 \end{bmatrix}.$$

Then

$$P_1 = \begin{bmatrix} 0.251 & 0.081 & 0.03 & 0.09 \\ 0.081 & 0.14 & -0.01 & 0.04 \\ 0.03 & -0.01 & 0.19 & -0.05 \\ 0.09 & 0.04 & -0.05 & 0.13 \end{bmatrix}, P_2 = \begin{bmatrix} 10.68 & -2.72 & -2.07 & -2.51 \\ -2.72 & 1.46 & 0.33 & 0.66 \\ -2.07 & 0.33 & 2.09 & 0.24 \\ -2.51 & 0.66 & 0.24 & 3.36 \end{bmatrix}.$$

The matrix L is calculated as

$$L = \begin{bmatrix} -1 & 2.126\\ 0.541 & -7.507 \end{bmatrix}.$$

Here  $l_{11} < 0$ ,  $l_{22} < 0$ ,  $l_{12} > 0$ ,  $l_{21} > 0$  and

$$l_{11} \cdot l_{22} - l_{12} \cdot l_{21} = 6.356834 > 0$$

and the conditions of Theorem 1 are satisfied.

$$k_1 = -\frac{l_{11}}{l_{12}} = 0.470366, \quad k_2 = -\frac{l_{21}}{l_{22}} = 0.072066$$

and the graphs of y = 0.470366x and y = 0.072066x are shown in Fig. 2.

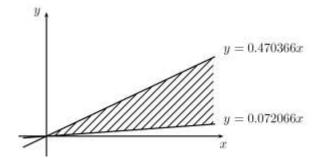


Figure 2.

 $w_1^0 = 1.515376, w_2^0 = 0.242416$  and by the above the matrix

$$P_* = 1.515376 P_1 + 0.242416 P_2$$

is a common solution.

#### **3. AN ALGORITHM FOR TWO COMPLEX MATRICES**

In this section we give an algorithm for a common P for given two Hurwitz stable complex matrices. For this purpose we establish the derivative formula for real functionals defined on the set of Hermitian matrices.

Let  $\mathcal{H}$  be the set of  $n \times n$ -dimensional Hermitian matrices. Assume that  $f: \mathcal{H} \to \mathbb{R}$  is a real matrix functional.

**Definition 1.** The functional f(P) defined on  $\mathcal{H}$  is called differentiable at the point P if there exists a Hermitian matrix  $\partial f|_P$  such that

$$f(P + \Delta P) = f(P) + \langle \partial f |_{P}, \Delta P \rangle + o(\Delta P),$$

where  $\frac{o(\Delta P)}{\|\Delta P\|} \to 0$  as  $\|\Delta P\| \to 0$ , where  $\|\cdot\|$  is the Frobenius norm. Here the inner product be defined by  $\langle \partial f|_P, \Delta P \rangle = tr(\partial f|_P, \Delta P)$ . A Hermitian  $n \times n$  matrix P has totally r = n(n+1)/2 entries:

$$P = \begin{bmatrix} x_1 & x_2 + jy_2 & \cdots & x_n + jy_n \\ x_2 - jy_2 & x_{n+1} & \cdots & x_{2n-1} + jy_{2n-1} \\ \vdots & \vdots & \ddots & \vdots \\ x_n - jy_n & x_{2n-1} - jy_{2n-1} & \cdots & x_r \end{bmatrix}.$$

Define real function g by

$$g(x_1, x_2, \dots, x_r, y_2, y_3, y_5, \dots, y_q) := f(P).$$

If the functional is differentiable at *P* then the function *g* is differentiable at  $(x_1, ..., x_r, y_2, ..., y_q)$ . We wish calculate the entries of the derivative matrix  $\partial f|_P$  in terms of the partial derivative of *g*.

**Theorem 2.** If the functional f(P) is differentiable at P, then

$$\partial f|_{P} = \begin{bmatrix} \frac{\partial g}{\partial x_{1}} & \frac{1}{2} \left( \frac{\partial g}{\partial x_{2}} + j \frac{\partial g}{\partial y_{2}} \right) & \cdots & \frac{1}{2} \left( \frac{\partial g}{\partial x_{n}} + j \frac{\partial g}{\partial y_{n}} \right) \\ \frac{1}{2} \left( \frac{\partial g}{\partial x_{2}} - j \frac{\partial g}{\partial y_{2}} \right) & \frac{\partial g}{\partial x_{n+1}} & \cdots & \frac{1}{2} \left( \frac{\partial g}{\partial x_{2n-1}} + j \frac{\partial g}{\partial y_{2n-1}} \right) \\ \vdots & \vdots & \ddots & \vdots \\ \frac{1}{2} \left( \frac{\partial g}{\partial x_{n}} - j \frac{\partial g}{\partial y_{n}} \right) & \frac{1}{2} \left( \frac{\partial g}{\partial x_{2n-1}} - j \frac{\partial g}{\partial y_{2n-1}} \right) & \cdots & \frac{\partial g}{\partial x_{r}} \end{bmatrix}.$$

**Proof.** To avoid cumbersome expressions we give the proof for the case n=3. For an arbitrary n the proof is identical.

Define  $3 \times 3$  Hermitian matrices

$$P = \begin{bmatrix} x_1 & x_2 + jy_2 & x_3 + jy_3 \\ x_2 - jy_2 & x_4 & x_5 + jy_5 \\ x_3 - jy_3 & x_5 - jy_5 & x_6 \end{bmatrix}$$

and

$$P + \Delta P = \begin{bmatrix} x_1 + \Delta x_1 & x_2 + \Delta x_2 + j(y_2 + \Delta y_2) & x_3 + \Delta x_3 + j(y_3 + \Delta y_3) \\ x_2 + \Delta x_2 - j(y_2 + \Delta y_2) & x_4 + \Delta x_4 & x_5 + \Delta x_5 + j(y_5 + \Delta y_5) \\ x_3 + \Delta x_3 - j(y_3 + \Delta y_3) & x_5 + \Delta x_5 - j(y_5 + \Delta y_5) & x_6 + \Delta x_6 \end{bmatrix}.$$

If f(P) is differentiable at P then the scalar function

$$g(x_1, x_2, x_3, x_4, x_5, x_6, y_2, y_3, y_5) = f(P)$$

is differentiable at the point  $u = (x_1, x_2, ..., x_6, y_2, y_3, y_5)$ . The following can be written:

$$g(u + \Delta u) = g(u) + \langle \partial g(u), \Delta u \rangle + o(\Delta u)$$
  
=  $g(u) + \frac{\partial g}{\partial x_1} \Delta x_1 + \frac{\partial g}{\partial x_2} \Delta x_2 + \dots + \frac{\partial g}{\partial x_6} \Delta x_6 + \frac{\partial g}{\partial y_2} \Delta y_2$   
+  $\frac{\partial g}{\partial y_3} \Delta y_3 + \frac{\partial g}{\partial y_5} \Delta y_5 + o(\Delta u)$ 

where  $\Delta u = (\Delta x_1, \Delta x_2, ..., \Delta x_6, \Delta y_2, \Delta y_3, \Delta y_5)$ , and  $\frac{o(\Delta u)}{\|\Delta u\|} \to 0$  as  $\|\Delta u\| \to 0$ .

Let the derivative matrix of f at P be

$$\partial f|_P = \begin{bmatrix} a_1 & a_2 + jb_2 & a_3 + jb_3 \\ a_2 - jb_2 & a_4 & a_5 + jb_5 \\ a_3 - jb_3 & a_5 - jb_5 & a_6 \end{bmatrix}.$$

We aim to express  $a_i$  (i = 1, ..., 6) and  $b_i$  (i = 2,3,5) in terms of the partial derivative of g. We have

$$tr(\partial f|_{P}.\Delta P) = a_{1}\Delta x_{1} + (a_{2} + jb_{2})(\Delta x_{2} - j\Delta y_{2}) + (a_{3} + jb_{3})(\Delta x_{3} - j\Delta y_{3}) + (a_{2} - jb_{2})(\Delta x_{2} + j\Delta y_{2}) + a_{4}\Delta x_{4} + (a_{5} + jb_{5})(\Delta x_{5} - j\Delta y_{5}) + (a_{3} - jb_{3})(\Delta x_{3} + j\Delta y_{3}) + (a_{5} - jb_{5})(\Delta x_{5} + j\Delta y_{5}) + a_{6}\Delta x_{6} = a_{1}\Delta x_{1} + 2a_{2}\Delta x_{2} + 2a_{3}\Delta x_{3} + a_{4}\Delta x_{4} + 2a_{5}\Delta x_{5} + a_{6}\Delta x_{6} + 2b_{2}\Delta y_{2} + 2b_{3}\Delta y_{3} + 2b_{5}\Delta y_{5}$$

on the other hand

$$<\partial g(u), \Delta u > = < \left(\frac{\partial g}{\partial x_1}, \frac{\partial g}{\partial x_2}, \dots, \frac{\partial g}{\partial y_5}\right), (\Delta x_1, \Delta x_2, \dots, \Delta y_5) >$$

$$= \frac{\partial g}{\partial x_1} \Delta x_1 + \frac{\partial g}{\partial x_2} \Delta x_2 + \frac{\partial g}{\partial x_3} \Delta x_3 + \frac{\partial g}{\partial x_4} \Delta x_4 + \frac{\partial g}{\partial x_5} \Delta x_5 + \frac{\partial g}{\partial x_6} \Delta x_6.$$

$$+ \frac{\partial g}{\partial y_2} \Delta y_2 + \frac{\partial g}{\partial y_3} \Delta y_3 + \frac{\partial g}{\partial y_5} \Delta y_5$$

If the functional f is differentiable at P

$$\operatorname{tr}(\partial f|_P.\Delta P) = <\partial g(u), \Delta u >$$

for all  $\Delta u = (\Delta x_1, \Delta x_2, ..., \Delta x_6, \Delta y_2, \Delta y_3, \Delta y_5)$ . Since  $\Delta x_i$  (i = 1, ..., 6),  $\Delta y_i$  (i = 2, 3, 5) are arbitrary then

$$a_{1} = \frac{\partial g}{\partial x_{1}}, \quad 2a_{2} = \frac{\partial g}{\partial x_{2}}, \quad 2a_{3} = \frac{\partial g}{\partial x_{3}}, \quad a_{4} = \frac{\partial g}{\partial x_{4}}, \quad 2a_{5} = \frac{\partial g}{\partial x_{5}}, \quad a_{6} = \frac{\partial g}{\partial x_{6}},$$
$$2b_{2} = \frac{\partial g}{\partial y_{2}}, \quad 2b_{3} = \frac{\partial g}{\partial y_{3}}, \quad 2b_{5} = \frac{\partial g}{\partial y_{5}}.$$

Therefore gradient of the functional f at P is

$$\partial f|_{P} = \begin{bmatrix} \frac{\partial g}{\partial x_{1}} & \frac{1}{2} \left( \frac{\partial g}{\partial x_{2}} + j \frac{\partial g}{\partial y_{2}} \right) & \frac{1}{2} \left( \frac{\partial g}{\partial x_{3}} + j \frac{\partial g}{\partial y_{3}} \right) \\ \frac{1}{2} \left( \frac{\partial g}{\partial x_{2}} - j \frac{\partial g}{\partial y_{2}} \right) & \frac{\partial g}{\partial x_{4}} & \frac{1}{2} \left( \frac{\partial g}{\partial x_{5}} + j \frac{\partial g}{\partial y_{5}} \right) \\ \frac{1}{2} \left( \frac{\partial g}{\partial x_{3}} - j \frac{\partial g}{\partial y_{3}} \right) & \frac{1}{2} \left( \frac{\partial g}{\partial x_{5}} - j \frac{\partial g}{\partial y_{5}} \right) & \frac{\partial g}{\partial x_{6}} \end{bmatrix}.$$

Let  $A_1$  and  $A_2$  be  $n \times n$  dimensional complex Hurwitz stable matrices.

**Theorem 3.** [6] Let  $\varepsilon > 0$  and *I* be the identity matrix. Then

$$\exists P > 0, \quad \begin{array}{l} A_1^*P + PA_1 < 0, \\ A_2^*P + PA_2 < 0, \end{array} \iff \quad \exists P > 0, \quad \begin{array}{l} A_1^*P + PA_1 + \varepsilon I \leq 0, \\ A_2^*P + PA_2 + \varepsilon I \leq 0. \end{array}$$

Let  $\varepsilon > 0$  and  $Q \ge 0$  be given. Denote by  $P_{A_2(Q)} > 0$  the solution of

$$A_2^*P + PA_2 = -Q - \varepsilon I$$

and define the following convex matrix functional:

$$F(Q) = \lambda_{\max} \Big( A_1^* P_{A_2(Q)} + P_{A_2(Q)} A_1 \Big).$$

If  $F(\tilde{Q}) < 0$  at some  $\tilde{Q} \ge 0$ , then the matrix  $P_{A_2(Q)}$  is a common solution.

Consider the minimization of F(Q). For Hermitian matrix H, the projection of H onto the convex cone of nonnegative definite matrices denote by  $[H]^+$ . For the convergence of the proposed algorithm we impose the following condition (see [6]).

**Condition 1.** For t > 0 there exists  $Q^* \ge 0$  such that the closed ball of radius t and centered at  $Q^*$  is contained in the set  $\{Q \ge 0: F(Q) < 0\}$ .

Consider the following convex problem

$$\begin{cases} F(Q) & \to & \min\\ Q \ge 0. \end{cases}$$

Using Theorem 2 we suggest the following solution algorithm.

#### Algorithm 1.

- 1. Choose  $Q^0 \ge 0$ . If  $F(Q^0) < 0$  then stop. Otherwise continue.
- 2. For k = 0, 1, 2, ... define

$$Q^{k+1} = \left[ Q^k - \mu_k \partial F(Q) \right]_{Q=Q^k}^{\dagger}$$

where

$$\mu_{k} \coloneqq \frac{\alpha F(Q^{k}) + t \left\| \partial F(Q) \right\|_{Q = Q^{k}}}{\left\| \partial F(Q) \right\|_{Q = Q^{k}}}$$

 $0 \le \alpha \le 2$  and t > 0 is defined from Condition 1.

3. If  $F(Q^k) < 0$  for some k then stop. The corresponding  $P_{A_2}(Q^k) < 0$  is a common solution.

Example 2. This example taken from [4]. Consider the following complex Hurwitz matrices

$$A_1 = \begin{bmatrix} -2 - j & 1 + 2j \\ -1.5 - j & -1.1 + j \end{bmatrix}, \quad A_2 = \begin{bmatrix} -2 - j & 3 + 4j \\ -2.5 - j & -2.5 + j \end{bmatrix}.$$

Let  $Q^0 = I$  and  $\varepsilon = 0.001$ . We have

$$P_{A_2(Q^0)} = \begin{bmatrix} 6.673 & -2.558 + 6.450j \\ -2.558 - 6.450j & 7.451 \end{bmatrix}$$

and  $F(Q^0) = 10.729 > 0$ . By Theorem 2,

$$\partial F(Q)|_{Q=Q^0} = \begin{bmatrix} 8.323 & -0.572 + 4.924j \\ -0.572 - 4.924j & 2.394 \end{bmatrix}$$

Therefore,

$$Q^{1} = \begin{bmatrix} Q^{0} - \mu_{0} \partial F(Q) |_{Q=Q^{0}} \end{bmatrix}^{+} = \begin{bmatrix} 0.270 & 0.055 - 0.472j \\ 0.055 + 0.472j & 0.839 \end{bmatrix}$$

where  $\alpha = 2, t = 1$  and  $\mu_0 = 0.262$ .

Algorithm 1 after 5 steps gives

$$Q^{5} = \begin{bmatrix} 0.318 & 0.044 - 0.564j \\ 0.044 + 0.564 & 1.006 \end{bmatrix}, \quad F(Q^{5}) = -0.085 < 0$$

and the matrix

$$P_{A_2(Q^5)} = \begin{bmatrix} 0.0582 & -0.001 - 0.046j \\ -0.001 + 0.046j & 0.126 \end{bmatrix}$$

is a common solution for  $A_1$  and  $A_2$ .

Example 3. Consider the following complex Hurwitz matrices

$$A_{1} = \begin{bmatrix} -3 & j & 1-j \\ 1+j & -3-3j & -2+j \\ 2-j & -3-2j & -4-j \end{bmatrix}, \quad A_{2} = \begin{bmatrix} -0.6+j & -2+3j & 1+2j \\ 3+j & -0.6-3j & 2+j \\ -2-j & -5-2j & -1.6-j \end{bmatrix}.$$

Let  $Q^0 = I$  and  $\varepsilon = 0.001$ . We have

$$P_{A_2(Q^0)} = \begin{bmatrix} 2.068 & -0.403 - 1.239j & 0.260 + 1.231j \\ -0.403 + 1.239j & 1.999 & -0.707 + 0.037j \\ 0.260 - 1.231j & -0.707 - 0.037j & 1.153 \end{bmatrix}$$

and  $F(Q^0) = 1.8808 > 0$ . By Theorem 1,

$$\partial F(Q)|_{Q=Q^0} = \begin{bmatrix} 0.403 & -0.171 + 0.236j & -0.816 + 0.277j \\ -0.171 - 0.236j & 0.368 & 0.094 + 0.805j \\ -0.816 - 0.277j & 0.094 - 0.805j & 1.106 \end{bmatrix}$$

Therefore,

$$Q^{1} = \begin{bmatrix} Q^{0} - \mu_{0} \partial F(Q) |_{Q=Q^{0}} \end{bmatrix}^{+}$$
  
= 
$$\begin{bmatrix} 0.834 & 0.217 + 0.031j & 0.473 - 0.230j \\ 0.217 - 0.031j & 0.836 & 0.011 - 0.499j \\ 0.473 + 0.230j & 0.011 + 0.499j & 0.574 \end{bmatrix}$$

where  $\alpha = 2, t = 1$  and  $\mu_0 = 1.308$ .

Algorithm 1 after 3 steps gives

$$Q^{3} = \begin{bmatrix} 0.813 & 0.277 + 0.060j & 0.508 - 0.428j \\ 0.277 - 0.060j & 0.875 & 0.188 - 0.652j \\ 0.508 + 0.428j & 0.188 + 0.652j & 0.828 \end{bmatrix}, \quad F(Q^{3}) = -0.08 < 0$$

and the matrix

$$P_{A_2(Q^3)} = \begin{bmatrix} 0.558 & 0.012 - 0.135j & 0.179 + 0.112j \\ 0.012 + 0.135j & 0.643 & -0.109 - 0.085j \\ 0.179 - 0.112j & -0.109 + 0.085j & 0.322 \end{bmatrix}$$

is a common solution for  $A_1$  and  $A_2$ .

#### **CONFLICT OF INTEREST**

The author stated that there are no conflicts of interest regarding the publication of this article.

## REFERENCES

- [1] Liberzon D. Switching in Systems and Control. Boston MA: Birkauser, 2003.
- [2] Lin H, Antsaklis PJ. Stability and Stabilizability of Switched Linear Systems: A Survey of Recent Results. IEEE Transactions on Automatic Control, 2009; 54(2): 308-322.
- [3] Shorten RN, Narendra KS. Necessary and sufficient conditions for the existence of a common quadratic Lyapunov function for a finite number of stable second order linear time-invariant systems. International Journal of Adaptive Control and Signal Processing, 2002; 16: 709-728.
- [4] Laffey TJ, Šmigoc H. Common solution to the Lyapunov equation for 2 × 2 complex matrices. Linear Algebra and its Applications, 2007; 420: 609-624.
- [5] Horn RA, Johnson CR. Matrix Analysis. Cambridge: Cambridge University Press, 1985.
- [6] Dzhafarov V, Büyükköroğlu T, Yılmaz Ş. On one application of convex optimization to stability of linear systems. Trudy Instituta Matematiki i Mekhaniki Ur0 Ran, 2015; 21(2): 320-328.
- [7] Liberzon D, Tempo R. Common Lyapunov functions and gradient algorithms. IEEE Transactions on Automatic Control, 2004; 49(6): 990-994.
- [8] Büyükköroğlu T, Esen Ö, Dzhafarov V. Common Lyapunov functions for some special classes of stable systems. IEEE Transactions on Automatic Control, 2011; 56(8): 1963-1967.