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# An application of guaranteed cost control to a 3-DOF model helicopter

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**Abstract:** This paper deals with guaranteed cost control for a model helicopter which has 3-DOF (degree of freedom): the elevation, pitch, and travel angles. One of main difficulties in designing a feedback controller for the helicopter is that the model includes nonlinearities. In this paper, these nonlinearities are considered as the uncertainty terms. Guaranteed cost control is applied not only to achieve the closed-loop stability but also to guarantee an adequate level of performance of the nonlinear 3-DOF model helicopter. A numerical example is shown to illustrate the effectiveness of the proposed method.

Keywords: guaranteed cost control, 3-DOF model helicopter, LMI

### **1 INTRODUCTION**

This paper deals with guaranteed cost control for a model helicopter which has 3-DOF (degree of freedom): the elevation, pitch, and travel angles [1], [2]. One of main difficulties in designing a feedback controller for the helicopter is that the dynamics include nonlinearities.

Although single-input single-output approaches have advantages in simple structure, straightforward and so on, these methods are difficult to consider uncertainties. Therefore, development of multi-input multi-output control approaches are widely applied, see e.g. [3], [4]. Moreover, avoiding difficulties in measurement of system state due to uncertainties, an observer can be applied to reconstruct the system dynamics [5].

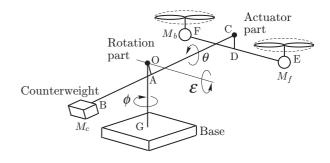
In this paper, these nonlinearities are considered as the uncertainty terms. By using Taylor's expansion, the state equation of a nonlinear 3-DOF model helicopter is changed to the form of a continuous-time uncertain system. Because the presence of the uncertainties may cause instability and bad performance on a controlled system, then guaranteed cost control method is applied.

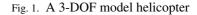
The objective of this paper is to propose a design method of guaranteed cost control with a minimal order observer for a 3-DOF model helicopter via linear matrix inequalities (LMIs) feasible solutions.

Finally, a numerical example is given to illustrate the effectiveness of the proposed method and it is shown that a 3-DOF nonlinear model helicopter can be stabilized by the guaranteed cost control method.

# 2 MODEL HELICOPTER

The dynamics of a 3-DOF model helicopter shown in Fig. 1. are described [1] as





$$\dot{\boldsymbol{x}}_{p} = \begin{bmatrix} p_{1}\cos\varepsilon + p_{2}\sin\varepsilon + p_{3}\dot{\varepsilon} + p_{4}\cos\theta v_{1} \\ p_{5}\cos\theta + p_{6}\sin\theta + p_{7}\dot{\theta} + p_{8}v_{2} \\ p_{9}\dot{\phi} + p_{10}\sin\theta v_{1} \\ \dot{\varepsilon} \\ \dot{\theta} \\ \dot{\phi} \end{bmatrix}$$
(1)

where  $p_i$ , (*i*=1,...,10) are model helicopter constants;  $\varepsilon$ ,  $\theta$ ,  $\phi$  are the elevation, pitch and travel angles and

$$egin{aligned} m{x}_p &= \dot{arepsilon}, \dot{eta}, arepsilon, eta, eta, eta, eta^T, \ v_1 &= V_f + V_b, \ v_2 &= V_f - V_b \end{aligned}$$

 $V_f$  and  $V_b$  are voltages applied to the front and rear motor, respectively.

#### **3 PROBLEM STATEMENT**

By using Taylor's expansion, a 3-DOF nonlinear model helicopter (1) can be expressed by the form

$$\dot{\boldsymbol{x}}(t) = (A + \Delta A(t))\boldsymbol{x}(t) + (B + \Delta B(t))\boldsymbol{u}(t) \quad (2)$$

$$(t) = C\boldsymbol{x}(t), \ C = [O \ I_3] \tag{3}$$

y

$$\boldsymbol{u}(t) = \begin{bmatrix} u_1 \\ u_2 \end{bmatrix}, \ u_1 = v_1 + \frac{p_1}{p_4}, \ u_2 = v_2 + \frac{p_5}{p_8}$$
(4)

where

$$\begin{split} &A = [a_{ij}], \ (i, j = 1, ..., 6), \\ &a_{11} = p_3, \ a_{14} = p_2, \ a_{22} = p_7, \ a_{25} = p_6, \\ &a_{33} = -\frac{p_1 p_{10}}{p_4}, \ a_{41} = a_{52} = a_{63} = 1, \\ &B = [b_{ij}], \ (i = 1, ..., 6, \ j = 1, 2), \\ &b_{11} = p_4, \ b_{21} = p_8, \\ &\Delta A = [\delta a_{ij}], \ (i, j = 1, ..., 6), \\ &\delta a_{11} = \Delta A_{11}, \ \delta a_{14} = \Delta A_{14}, \ \delta a_{15} = \Delta A_{15}, \\ &\delta a_{22} = \Delta A_{22}, \ \delta a_{25} = \Delta A_{25}, \ \delta a_{33} = \Delta A_{33}, \\ &\delta a_{35} = \Delta A_{35}, \\ &\Delta B = [\delta b_{ij}], \ (i = 1, ..., 6, \ j = 1, 2), \\ &\delta b_{11} = \Delta B_{11}, \ \delta b_{22} = \Delta B_{22}, \ \delta b_{31} = \Delta B_{31}, \\ &\Delta A_{14} = p_2(-\frac{1}{3!}x_4^2 + O(x_4^4)) + p_1(-\frac{1}{2}x_4 + O(x_4^3)) + \Delta p_2, \\ &\Delta A_{15} = p_1(-\frac{1}{2}x_5 + O(x_5^3)), \ \Delta A_{22} = \Delta p_7, \\ &\Delta A_{25} = p_5(-\frac{1}{2}x_5 + O(x_5^3)) + p_6(-\frac{1}{3!}x_6^2 + O(x_6^4)) + \Delta p_6, \\ &\Delta A_{33} = \Delta p_9, \\ &\Delta A_{35} = -\frac{p_1 p_{10}}{p_4}(-\frac{1}{3!}x_5^2 + O(x_5^4)) - \Delta(p_1 p_{10}/p_4), \\ &\Delta B_{11} = p_4(-\frac{1}{2}x_5 + O(x_5^4)) + \Delta p_4, \\ &\Delta B_{22} = \Delta p_8, \ \Delta B_{31} = p_{10}(x_5 - \frac{1}{3!}x_5^3 + O(x_5^5)), \end{split}$$

while other elements are zero, and  $\boldsymbol{x}(t) \in \Re^n$  is the state vector,  $\boldsymbol{u}(t) \in \Re^r$  is the control input vector,  $\boldsymbol{y}(t) \in \Re^m$  is the measured output vector,  $\Delta p_i$  denote uncertain terms of  $p_i$ , A, B, C are known constant real-valued matrices with appropriate dimensions. Due to the system constraint, we have the bounds of  $|\varepsilon|$ ,  $|\theta|$  and  $|\phi|$  as  $\varepsilon_{max}(=0.3)$ ,  $\theta_{max}(=0.3)$ .

Under the limitations of  $|\varepsilon|$ ,  $|\theta|$  and  $|\phi|$ , matrices  $\Delta A(t)$  and  $\Delta B(t)$  can be represented as

$$\Delta A(t) = D_A F_A(t) E_A, \ \Delta B(t) = D_B F_B(t) E_B$$
 (5)

with

$$F_A^T(t)F_A(t) \le I, \ F_B^T(t)F_B(t) \le I$$

where  $D_A$ ,  $D_B$ ,  $E_A$ ,  $E_B$  are constant real-valued known matrices with appropriate dimensions, and  $F_A(t)$  and  $F_B(t)$  are real time-varying unknown continuous and deterministic matrices.

We assume that the initial state variable x(0) is unknown, but their mean and covariance are known

$$E[\boldsymbol{x}(0)] = \boldsymbol{m}_0 \qquad (6)$$
$$E (\boldsymbol{x}(0) - \boldsymbol{m}_0)(\boldsymbol{x}(0) - \boldsymbol{m}_0)^T = \Sigma_0 > O \qquad (7)$$

where  $E[\cdot]$  denotes the expected value operator.

The problem considered here is to design a guaranteed cost controller with a minimal order observer so as to achieve an upper bound on the following quadratic performance index

$$E[J] = E\left[\int_0^\infty (\boldsymbol{x}^T(t)Q\boldsymbol{x}(t) + \boldsymbol{u}^T(t)R\boldsymbol{u}(t))dt\right] < E[J^*]$$
(8)

associated with the uncertain system (2) where Q and R are given symmetric positive-definite matrices.

#### 4 GUARANTEED COST CONTROLLER DE-

# SIGN

Design of a guaranteed cost controller is described in the following equations. Here, a minimal order observer is given by

$$\dot{\boldsymbol{z}}(t) = D\boldsymbol{z}(t) + E\boldsymbol{y}(t) + F\boldsymbol{u}(t)$$
(9)

$$\hat{\boldsymbol{x}}(t) = P\boldsymbol{z}(t) + W\boldsymbol{y}(t)$$
(10)

with

$$D = A_{11} + LA_{21}, PT + WC = I_6,$$
  

$$F = TB, TA - DT = EC, A = \left[\frac{A_{11} | A_{12}}{A_{21} | A_{22}}\right],$$
  

$$P = I_3 \mathbf{0}^{T}, T = I_3 L$$

and a controller is assumed to have a form of

$$\boldsymbol{u}(t) = K\hat{\boldsymbol{x}}(t), \ K = -R^{-1}B^T S_1 \tag{11}$$

where  $S_1$  is a symmetric positive definite matrix.

Then, the following Theorem gives a design method of guaranteed cost control to the 3-DOF model helicopter (2)-(3).

*Theorem 1.* If the following matrix inequalities optimization problem; min  $\{\gamma_0 + \gamma_1 + \gamma_2 + \gamma_3 + \gamma_4\}$  subject to

$$\begin{bmatrix} \Lambda_0 & X^T & XE_A & XE_A \\ * & -Q^{-1} & 0 & 0 \\ * & 0 & -\alpha_1 I & 0 \\ * & 0 & 0 & -\alpha_4 I \end{bmatrix} < 0$$
(12)

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$$\begin{bmatrix} \bar{\Lambda}_{0} G_{1} & G_{2} & G_{2} & G_{3} & G_{3} & G_{4} \\ * & -R & 0 & 0 & 0 & 0 \\ * & 0 & -\alpha_{5,inv}I & 0 & 0 & 0 \\ * & 0 & 0 & -\alpha_{6,inv}I & 0 & 0 & 0 \\ * & 0 & 0 & 0 & -\alpha_{3}I & 0 & 0 \\ * & 0 & 0 & 0 & 0 & -\alpha_{6}I & 0 \\ * & 0 & 0 & 0 & 0 & 0 & -\alpha_{4,inv}I \end{bmatrix} < 0$$

$$(13)$$

$$\sum_{k=1}^{6} e_{6k}^{T} \Theta_{0} e_{6k} < \gamma_{0}, \sum_{k=1}^{3} e_{3k}^{T} \Theta_{1} e_{3k} < \gamma_{1},$$

$$\sum_{k=1}^{3} e_{3k}^{T} \Theta_{2} e_{3k} < \gamma_{2}, \sum_{k=1}^{3} e_{3k}^{T} \Theta_{3} e_{3k} < \gamma_{3}$$
(14)
$$\begin{bmatrix} -\gamma_{4} v_{1}^{T} Y^{T} v_{2}^{T} Y^{T} v_{3}^{T} Y^{T} \\ Y v_{1} - S_{2} \\ Y v_{3} & -S_{2} \end{bmatrix} < 0$$
(15)

where

$$\begin{split} \Lambda_{0} &= AX + XA^{T} - BR^{-1}B^{T} + (\alpha_{2} + \alpha_{3})D_{B}D_{B}^{T} \\ &+ \alpha_{1}D_{A}D_{A}^{T} + (\alpha_{2,inv} + \alpha_{5,inv})BR^{-1}E_{B}^{T}E_{B}R^{-1}B^{T} \\ \bar{\Lambda}_{0} &= S_{2}A_{11} + A_{11}^{T}S_{2} + YA_{21} + A_{21}^{T}Y^{T}, \\ Y &= S_{2}L, \ Z &= [S_{2} \ Y], \ G_{1} &= P^{T}S_{1}B, \\ G_{2} &= ZD_{B}, \ G_{3} &= P^{T}S_{1}BR^{-1}E_{B}^{T}, \ G_{4} &= ZD_{A}, \\ \Theta_{0} &= \frac{1}{2}(S_{1}(\Sigma_{0} + m_{0}m_{0}^{T}) + (\Sigma_{0} + m_{0}m_{0}^{T})^{T}S_{1}), \\ \Theta_{1} &= \frac{1}{2}(S_{2}\Sigma_{11} + \Sigma_{11}S_{2}), \ \Theta_{2} &= \frac{1}{2}(Y\Sigma_{21} + \Sigma_{21}^{T}Y^{T}), \\ \Theta_{3} &= \frac{1}{2}(Y^{T}\Sigma_{12} + \Sigma_{12}^{T}Y), \ e_{ik} &= \ \mathbf{0}_{k-1}^{T} \ \mathbf{1} \ \mathbf{0}_{i-k}^{T} \ ^{T}, \\ \Sigma_{0} &= \begin{bmatrix} \Sigma_{11} \ \Sigma_{12} \\ \Sigma_{21} \ \Sigma_{22} \end{bmatrix}, \ \Sigma_{22}^{1/2} &= [\mathbf{v}_{1}, \mathbf{v}_{2}, \mathbf{v}_{3}] \end{split}$$

has a solution  $S_1 > 0$ ,  $S_2 > 0$ , X > 0, Y, Z,  $\alpha_1 > 0$ ,  $\alpha_2 > 0$ ,  $\alpha_{2,inv} > 0$ ,  $\alpha_3 > 0$ ,  $\alpha_4 > 0$ ,  $\alpha_{4,inv} > 0$ ,  $\alpha_{5,inv} > 0$ ,  $\alpha_6 > 0$ ,  $\alpha_{6,inv} > 0$ ,  $\gamma_0$ ,  $\gamma_1$ ,  $\gamma_2$ ,  $\gamma_3$ ,  $\gamma_4$  which satisfy the relation  $\alpha_2^{-1} = \alpha_{2,inv}$ ,  $\alpha_4^{-1} = \alpha_{4,inv}$ ,  $\alpha_6^{-1} = \alpha_{6,inv}$  and  $S_1^{-1} = X$ , then the minimal order observer-based control law (9)-(11) is a guaranteed cost controller which gives the minimum expected value of the guaranteed cost

$$E[J^*] = E\left[\boldsymbol{x}^T(0)S_1\boldsymbol{x}(0) + \boldsymbol{\xi}^T(0)S_2\boldsymbol{\xi}(0)\right] \quad (16)$$

where  $\boldsymbol{\xi}(t) = \boldsymbol{z}(t) - T\boldsymbol{x}(t)$  is the estimated error of the minimal order observer.

*Remark 1:* Since inequalities in (12) and (13) contain scalars and matrices that satisfy inverse relations  $S_1^{-1} = X$ ,  $\alpha_2^{-1} =$ 

 $\alpha_{2,inv}, \alpha_4^{-1} = \alpha_{4,inv}$ , and  $\alpha_6^{-1} = \alpha_{6,inv}$ . then an iterative LMI algorithm is adopted to solve [6],[7].

#### **5 A NUMERICAL EXAMPLE**

The nominal values of the model helicopter are as follows:

$$\begin{split} p_1 &= [-(M_f + M_b)gL_a + M_cgL_c]/J_{\varepsilon}, \\ p_2 &= [-(M_f + M_b)gL_a \tan\delta_a + M_cgL_c \tan\delta_c]/J_{\varepsilon}, \\ p_3 &= \eta_{\varepsilon}/J_{\varepsilon}, \ p_4 = K_mL_a/J_{\varepsilon}, \ p_5 = (-M_f + M_b)gL_h/J_{\theta}, \\ p_6 &= -(M_f + M_b)gL_h \tan\delta_h/J_{\theta}, \ p_7 = -\eta_{\theta}/J_{\theta}, \\ p_8 &= K_mL_h/J_{\theta}, \ p_9 &= -\eta_{\phi}/J_{\phi}, p_{10} = -K_mL_a/J_{\phi}, \\ \delta_a &= \tan^{-1}[(L_d + L_e)/L_a], \ \delta_c &= \tan^{-1}(L_d/L_c), \\ \delta_h &= \tan^{-1}(L_e/L_h), \ J_{\varepsilon} &= 0.86 \text{ kg m}^2, J_{\theta} &= 0.044 \text{ kg m}^2, \\ J_{\phi} &= 0.82 \text{ kg m}^2, L_a &= 0.62 \text{ m}, \ L_c &= 0.44 \text{ m}, \\ L_d &= 0.05 \text{ m}, \ L_e &= 0.02 \text{ m}, \ L_h &= 0.177 \text{ m}, \\ M_f &= 0.69 \text{ kg}, \ M_b &= 0.69 \text{ kg}, \ M_c &= 1.67 \text{ kg}, \\ K_m &= 0.5 \text{ N/V}, \ g &= 9.81 \text{ m/s}^2, \\ \eta_{\varepsilon} &= 0.001 \text{ kg m}^2/\text{s}, \ \eta_{\theta} &= 0.001 \text{ kg m}^2/\text{s}, \\ \eta_{\phi} &= 0.005 \text{ kg m}^2/\text{s}, \end{split}$$

and the uncertain parameters  $\Delta p_2$ ,  $\Delta p_3$ ,  $\Delta p_4$ ,  $\Delta p_5$ ,  $\Delta p_6$ ,  $\Delta p_7$ ,  $\Delta p_8$ ,  $\Delta p_9$ ,  $\Delta (p_1 p_{10}/p_4)$  are 5% of each  $p_2, p_3$ ,  $p_4$ ,  $p_5$ ,  $p_6$ ,  $p_7$ ,  $p_8$ ,  $p_9$  and  $(p_1 p_{10}/p_4)$ , respectively. Next,  $D_A$ ,  $E_A$ ,  $D_B$ ,  $E_B$ ,  $m_0$ ,  $\Sigma_0$ , R, Q are given as

$$\begin{split} D_{A} &= d_{A_{ij}} \ , (i, j = 1, ..., 6) \\ d_{A_{11}} &= \sqrt{|\Delta A_{11}|}, d_{A_{14}} = \sqrt{|\Delta A_{14}|}, \\ d_{A_{15}} &= \sqrt{|\Delta A_{15}|}, d_{A_{22}} = \sqrt{|\Delta A_{22}|}, \\ d_{A_{26}} &= -\sqrt{|\Delta A_{25}|}, d_{A_{33}} = -\sqrt{|\Delta A_{35}|}, \\ F_{A}(t) &= I_{6}, E_{A} = e_{A_{ij}} \ , (i, j = 1, ..., 6) \\ e_{A_{11}} &= \sqrt{|\Delta A_{11}|}, e_{A_{22}} = \sqrt{|\Delta A_{22}|}, \\ e_{A_{33}} &= \frac{\Delta A_{33}}{\sqrt{|\Delta A_{35}|}}, e_{A_{35}} = \sqrt{|\Delta A_{35}|}, \\ e_{A_{44}} &= \sqrt{|\Delta A_{14}|}, e_{A_{55}} = \sqrt{|\Delta A_{15}|}, \\ e_{A_{65}} &= \sqrt{|\Delta A_{25}|} \\ D_{B} &= d_{B_{ij}} \ , (i = 1, ..., 6; j = 1, 2) \\ d_{B_{12}} &= \frac{-\Delta B_{11}}{\sqrt{|\Delta B_{31}|}}, d_{B_{21}} = \sqrt{|\Delta B_{22}|}, \\ d_{B_{22}} &= -\frac{\sqrt{|\Delta B_{22}|} \times \sqrt{|\Delta B_{11}|}}{\sqrt{|\Delta B_{31}|}}, d_{B_{32}} = \sqrt{|\Delta B_{31}|} \\ F_{B}(t) &= I_{2}, E_{B} = \begin{bmatrix} -\sqrt{|\Delta B_{11}|} \\ -\sqrt{|\Delta B_{31}|} \\ 0 \end{bmatrix}, \\ \mathbf{m}_{0} &= \mathbf{0}_{6}, \Sigma_{0} = 0.036I_{6}, R = I_{2}, \\ Q &= \operatorname{diag}(0.1, 0.1, 0.1, 1, 1, 1), \end{split}$$

while other elements of  $d_{A_{ij}}$ ,  $e_{A_{ij}}$ , and  $d_{B_{ij}}$  are zero.

Results of the controller gain K, the observer gain L and the expected guaranteed cost  $E[J^*]$  are obtained below

$$K = \begin{bmatrix} -2.4649 & 0.0078 & 0.0112 & -0.1667 & -0.0467 & -0.0033 \\ 0.0433 & -2.4787 & 7.2135 & -0.1399 & -3.0527 & 2.3029 \end{bmatrix}$$
$$L = \begin{bmatrix} -2.9302 & 0.0574 & -0.1278 \\ 0.0574 & -2.3582 & 2.6908 \\ -0.1278 & 2.6908 & -7.6101 \end{bmatrix}, E[J^*] = 28.9098.$$

Figure 2 shows the transition of the guaranteed cost, and Fig. 3-5 show the trajectories of elevation, pitch and travel angles with  $\boldsymbol{x}(0) = [0 \ 0 \ 0 \ 0.2 \ 0.2 \ 0.1]^T$ .

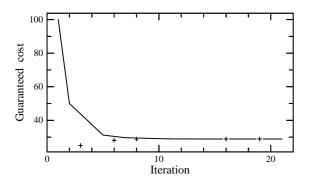


Fig. 2. Transition of the guaranteed cost

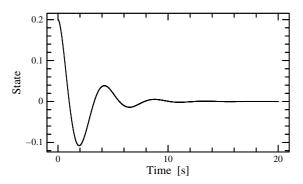


Fig. 3. Trajectory of elevation angle  $\varepsilon$ 

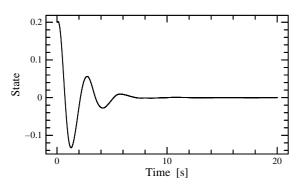


Fig. 4. Trajectory of pitch angle  $\theta$ 

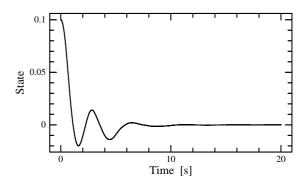


Fig. 5. Trajectory of travel angle  $\phi$ 

#### 6 CONCLUSION

This paper discusses a design method of guaranteed cost control with a minimal order observer for a 3-DOF nonlinear model helicopter via linear matrix inequalities (LMIs). The results show the effectiveness of the proposed method.

# REFERENCES

- Ishitobi M, Nishi M, Nakasaki K (2010), Nonlinear adaptive model following control for a 3-DOF tandem-rotor model helicopter. *Control Eng. Practice* 18(8):936-943
- [2] Kutay AT, Calise AJ, Idan M, Hovakimyan N (2005), Experimental results on adaptive output feedback control using a laboratory model helicopter. *IEEE Trans.* on Control Syst. Technology 13(2):196-202
- [3] Mahony R, Hamel T (2004), Robust trajectory tracking for a scale model autonomous helicopter. *Int. J. Robust* & *Nonlinear Control* 14(2):1035-1059
- [4] Marconi L, Naldi R (2007), Robust full degree-offreedom tracking control of a helicopter. *Automatica* 43(11):1909-1920
- [5] Lien CH (2005), Guaranteed cost observer-based controls for a class of uncertain neutral time-delay systems.
   J. Optim. Theory & Appl. 126(1):137-156
- [6] Matsunaga D, Ishitobi M, Kunimatsu S (2009), An ILMI approach to guaranteed cost controllers with a minimal order observer. *Proc. ICROS-SICE International Joint Conference*:583-588
- [7] Susanto E, Ishitobi M, Kunimatsu S, Matsunaga D (2011), A minimal order observer-based guaranteed cost controller for uncertain time-varying delay systems. *IMA J. Math. Control Inf.*, to appear