

DYNAMIC FOR DUAL-ARM FLOATING SPACE ROBOT WITH CLOSED-CHAIN AND RECURRENT ROBUST FUZZY NEURAL NETWORK FOR OBJECT GRASPING

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ABSTRACT

The dynamic model and the control method for dual-arm space robot grasping a target are discussed. Based on the Lagrange method, the dynamic model of an open chain space robot is established. The dynamics equations of dual-arm space robot with closed-chain are derived by incorporating the closed-loop constraints equations into equation of open chain system. Upon this basis, the recurrent robust fuzzy neural network control is proposed for the closed-chain system with uncertain parameters to complete the accurate control of load position and attitude. The recurrent fuzzy neural network is used to approximate the unknown part with H_∞ tracking characteristic to overcome the effects caused by system parameter perturbation. At last, numerical examples simulate the process of a planar dual-arm space robot transferring a target and verify the efficiency of the control scheme.

1 INTRODUCTION

With the development of space technology, there has been much interest in the space robot^[1-2] recently. The space robot system has been used in the space missions such as the surviving, construction, repair and maintenance of satellites and space stations. The free-floating space is a special kind of space manipulator which the base is free under the micro-gravity space environment. For the dynamic coupling between the manipulator and the base of the free floating space manipulator, the trajectory control of a space robot is more complicated than that of a terrestrial robotic system, and the situation will be worse when some of system inertial parameters are uncertain^[3]. Walker et al.^[4] proposed adaptive control for space robot with uncontrolled position and attitude of base. Chen et al.^[5] proposed robust control methods for a space robot system with uncertain parameters were estimated on-line. Over the last ten years, many control scheme studies has been applied in space robot, such as adaptive fuzzy control^[6], neural network control^[7], etc. Most of the space robot researches mainly focus on the control of single arm space robot, the control of dual-arm

space robot research is less. And none of the aforementioned studies has been performed against the closed-loop constraints of a dual-arm space robot. This is because the closed chain dual-arm space robot has some problems: (1) the closed chain grasping system introduces closed-loop constraint; (2) closed chain system controllers are redundancy and need distribution torques reasonably.

Since the heavy load and task complexity is growing up in space mission, multi-arm systems become more and more important in future. The closed-loop configuration will be formed when the manipulator arms grasp a common target. The closed-loop constraints imposes both kinematic and dynamic problem on the space robot system. Because the aforementioned problem, the control scheme design for dual-arm space robot grasping a target will be more complicated than that for the unconstrained one. Dual-arm space robot have been applied some effective works on the tracking control with open chain^[8]. Wang et al.^[9] developed a robust control method for closed chain dual-arm space robot; however, the system need feedback force information of end-effector.

This paper studies the tracking control scheme of a dual-arm space robot to manipulate a target. The dynamic equations of open chain dual-arm space robot are derived by applying the Lagrange equations. With the closed-loop constraints equations, the reduced-order equations of the dual-arm space robot with closed-chain are obtained. Based on the dynamic model, a recurrent robust fuzzy neural network control is proposed for control the position and attitude of the base and load. The uncertain part is approximated by neural network on-line, and H_∞ robust item is used to overcome the approximation error. Global asymptotic stability is proven by using Lyapunov's method. The numerical simulations show the effectiveness of the proposed control scheme.

2 SYSTEM DYNAMICS AND KINEMATICS ANALYSIS OF DUAL-ARM SYSTEM

The planar space robot consists of a free-floating base and two rigid manipulators, as shown in Fig1. Choosing origin O is located at an arbitrary point in planar, the inertia coordinate inertial coordinate frames XOY is built. Choosing the local frame coordinate of base and each link x_i, y_i ($i=1,2,\dots,6$). O_0 、 O_m are the mass center of base and load. O_i ($i=1,2,\dots,6$) is the mass center of each link. O_m is on the line connection between O_3 and O_6 . l_i ($i=1,2,\dots,6$), the length of $O_0 O_1$ and $O_0 O_4$ are d_0 . The distance of mass center of load O_m to end-effector are d_L and d_R . The dual-arm space robot forms a closed-chain after the capture operation.

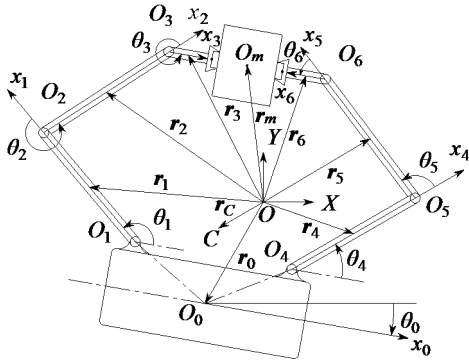


Figure1: Structure of dual-arm space robot with closed chain

2.1 Kinematics and dynamics analysis of open chain system

The dual-arm space robot with closed-chain is virtual disconnect at point b , and the virtual unconstrained system is shown in Fig1. I_i is the rotational inertia of each part, m_i is quality of each part. x_0 、 y_0 are mass center coordinates of base.

The generalized coordinates matrix of open chain system is defined as:

$$\mathbf{q} = [x_0 \quad y_0 \quad \theta_0 \quad \theta_L^T \quad \theta_R^T]^T \quad (1)$$

where $\theta_L = [\theta_1 \quad \theta_2 \quad \theta_3]^T$ denotes corners of left arm, $\theta_R = [\theta_4 \quad \theta_5 \quad \theta_6]^T$ denotes corners of left arm, θ_0 is the attitude of base, system has nine degrees of freedom.

The dynamic equations of dual-arm space robot with open chain is obtained by the Lagrange equations

$$\mathbf{M} \begin{pmatrix} \ddot{x}_0 \\ \ddot{y}_0 \\ \ddot{\theta}_0 \\ \ddot{\theta}_L \\ \ddot{\theta}_R \end{pmatrix} + \mathbf{H} \begin{pmatrix} \dot{x}_0 \\ \dot{y}_0 \\ \dot{\theta}_0 \\ \dot{\theta}_L \\ \dot{\theta}_R \end{pmatrix} = \boldsymbol{\tau} \quad (2)$$

$\boldsymbol{\tau}$ is the generalized control torque, For the convenience of description, system dynamic equations

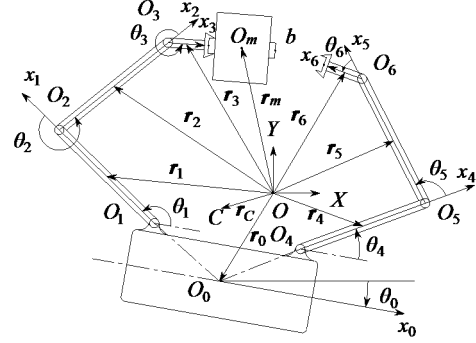


Figure2: Structure of unconstrained dual-arm space robot

is rewritten as:

$$\mathbf{M}\ddot{\mathbf{q}} + \mathbf{H}\dot{\mathbf{q}} = \boldsymbol{\tau} \quad (3)$$

where $\mathbf{M} \in \mathbf{R}^{9 \times 9}$ a symmetric positive definite system inertia matrix. $\mathbf{H}\dot{\mathbf{q}} \in \mathbf{R}^{9 \times 1}$ is a vector of the centripetal and coriolis factors. For arbitrary $\mathbf{x} \in \mathbf{R}^{9 \times 1}$, it satisfies that^[10]:

$$\mathbf{x}^T (\dot{\mathbf{M}} - 2\mathbf{H})\mathbf{x} = 0 \quad (4)$$

$\dot{\mathbf{M}} - 2\mathbf{H}$ is skew symmetric.

2.2 Kinematics and dynamics analysis of closed-chain system

The dynamic equations of reduced-order form for the closed-chain system will be established in this section. Selecting a motion reference point b at the cutting surface. The motion reference points have same position and speed in inertial frame and body-fixed coordinate system.

According to open chain system of geometric relations, the position vectors from motion reference points to O_0 are given as:

$$\begin{cases} \mathbf{r}_{0L} = (x_{0L} \quad y_{0L}) \\ \mathbf{r}_{bL} = \mathbf{r}_{0L} + l_1 \mathbf{e}_{01} + l_2 \mathbf{e}_{02} + (d_3 + d_L + d_R) \mathbf{e}_{03} \\ \mathbf{r}_{0R} = (x_{0R} \quad y_{0R}) \\ \mathbf{r}_{bR} = \mathbf{r}_{0R} + l_4 \mathbf{e}_{04} + l_5 \mathbf{e}_{05} + d_6 \mathbf{e}_{06} \end{cases} \quad (5)$$

\mathbf{r}_{0L} and \mathbf{r}_{0R} are position vectors from O_1 and O_4 to O_0 in base-fix coordinate, \mathbf{e}_{0i} is base vector.

Depending on the characters of motion reference points, thus

$$\mathbf{N}_L \dot{\boldsymbol{\theta}}_L = \mathbf{N}_R \dot{\boldsymbol{\theta}}_R \quad (6)$$

Let

$$\mathbf{U} = [\mathbf{I}_{6 \times 6} \quad \mathbf{U}_1^T] \quad (7)$$

$$U_1 = [\mathbf{O}_{3 \times 3} \quad \mathbf{N}_R^{-1} \mathbf{N}_L] \quad (8)$$

$\mathbf{I}_{n \times n}$ is a $n \times n$ unit matrix, $\mathbf{O}_{n \times n}$ is a $n \times n$ zero matrix.

The generalized coordinates matrix of closed-chain system is defined as:

$$\mathbf{q}_f = [x_0 \quad y_0 \quad \theta_0 \quad \boldsymbol{\theta}_L^T] \quad (9)$$

According to Eq.(9), Eq. (10), Eq. (11)and Eq. (12), can get:

$$\dot{\mathbf{q}} = \mathbf{U} \dot{\mathbf{q}}_f \quad (10)$$

Differentiating Eq. (10), we have

$$\ddot{\mathbf{q}} = \mathbf{U}^T \ddot{\mathbf{q}}_f + \dot{\mathbf{U}}^T \dot{\mathbf{q}}_f \quad (11)$$

Substituting Eq.(10), Eq.(11) into Eq.(3), the dynamic equations of closed-chain are obtained as follow:

$$\bar{\mathbf{M}} \ddot{\mathbf{q}}_f + \bar{\mathbf{H}} \dot{\mathbf{q}}_f = \bar{\boldsymbol{\tau}} \quad (12)$$

where

$$\bar{\mathbf{M}} = \mathbf{U} \mathbf{M} \mathbf{U}^T \quad (13)$$

$$\bar{\mathbf{H}} = \mathbf{U} [\mathbf{H} \mathbf{U}^T + \mathbf{M} \dot{\mathbf{U}}^T] \quad (14)$$

$$\bar{\boldsymbol{\tau}} = \mathbf{U} \boldsymbol{\tau} \quad (15)$$

In the dynamic equations of closed-chain system, For arbitrary $\mathbf{y} \in \mathbf{R}^{6 \times 1}$, $\bar{\mathbf{M}} \in \mathbf{R}^{6 \times 6}$ 、 $\bar{\mathbf{H}} \in \mathbf{R}^{6 \times 6}$ still satisfy that:

$$\mathbf{y}^T (\dot{\bar{\mathbf{M}}} - 2\bar{\mathbf{H}}) \mathbf{y} = 0 \quad (16)$$

$\dot{\bar{\mathbf{M}}} - 2\bar{\mathbf{H}}$ is skew symmetric.

So far, we have obtain the dynamic equations of closed-chain system.

3 RECURRENT ROBUST FUZZY NEURAL NETWORK CONTROL

3.1 Recurrent robust fuzzy neural network

The four-layer recurrent fuzzy neural network^[11] is shown in fig2.

Input layer

$$In_i^{(1)} = Out_i^{(1)} = x_i \quad (i = 1, 2, \dots, m) \quad (17)$$

Superscript means the which layer the input and output belong to.

Member function layer:

$$Out_{ij}^{(2)} = \mu(Out_i^{(1)}) = e^{-((Out_i^{(1)} - a_j^2) / b_j^2)} \quad (18)$$

Fuzzy reasoning layer:

$$Out_k^{(3)} = \prod w_{jk} In_j^{(3)} w_k Out_{k-1}^{(3)} \quad (k = 1, 2, \dots, l) \quad (19)$$

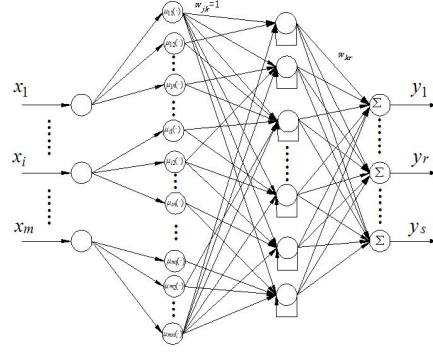


Figure3: Structure of recurrent fuzzy neural network

Output layer, the layer of defuzzification:

$$y_r = Out_r^{(4)} = w_{kr} In_k^{(4)} \quad (r = 1, 2, \dots, s) \quad (20)$$

Output can be written as follow:

$$\mathbf{u} = \mathbf{W}_r^T \boldsymbol{\Phi} \quad (21)$$

3.2 Control design of closed chain system

In order to control the position and attitude of both the base and the load, the output matrix of system is defined as:

$$\mathbf{X} = [x_0 \quad y_0 \quad \theta_0 \quad x_m \quad y_m \quad \theta_m]^T \quad (22)$$

Differentiating the Eq.(22):

$$\dot{\mathbf{X}} = \mathbf{J}_0(\mathbf{q}_f) \dot{\mathbf{q}}_f \quad (23)$$

$\mathbf{J}_0(\mathbf{q}_f) \in \mathbf{R}^{6 \times 6}$ is the Jacobian matrix.

Defining the output error as follow:

$$\mathbf{e} = \mathbf{X}_d - \mathbf{X} \quad (24)$$

With Eq.(24), the dynamic equations of closed-chain system can be rewritten as:

$$\mathbf{M}_X \ddot{\mathbf{X}} + \mathbf{H}_X \dot{\mathbf{X}} = \bar{\boldsymbol{\tau}} \quad (25)$$

wherer $\mathbf{M}_X = \bar{\mathbf{M}} \mathbf{J}_0^{-1}$, $\mathbf{H}_X = \bar{\mathbf{H}} \mathbf{J}_0^{-1} - \bar{\mathbf{M}} \dot{\mathbf{J}}_0^{-1} \mathbf{J}_0^{-1}$.

Since the uncertain parameters of system, we can get:

$$\hat{\mathbf{M}}_X \ddot{\mathbf{X}} + \hat{\mathbf{H}}_X \dot{\mathbf{X}} + \mathbf{f}(\boldsymbol{\delta}) = \bar{\boldsymbol{\tau}} \quad (26)$$

where $\hat{\mathbf{M}}_X$ 、 $\hat{\mathbf{H}}_X$ is nominal model, $\Delta \mathbf{M}_X$ 、 $\Delta \mathbf{H}_X$ is model error, and $\mathbf{f}(\boldsymbol{\delta}) = \Delta \mathbf{M}_X \ddot{\mathbf{X}} + \Delta \mathbf{H}_X \dot{\mathbf{X}}$.

Using recurrent fuzzy neural network to approximate

the unknown part $f(\delta)$ as follow:

$$f(\delta) = \mathbf{W}^{*T} \Phi + \Delta\delta \quad (27)$$

where \mathbf{W}^* is the optimal weight matrix. Φ is a column correlation with primary function, $\Delta\delta$ is approximate error.

For the uncertain parameters space robot, design the control as follow:

$$\tau = \hat{M}_X(\ddot{X}_d + K_v \dot{e} + K_p e) + \hat{H}_X \dot{X} + \mathbf{W}^T \Phi + \mu \quad (28)$$

μ is Robust control term, which is used to eliminate the approximate error.

With Eq.(28) and Eq.(25), we have:

$$\hat{M}_X(\ddot{e} + K_v \dot{e} + K_p e) - \tilde{W}^T \Phi - \Delta\delta + \mu = 0 \quad (29)$$

where $\tilde{W} = \mathbf{W}^* - \mathbf{W}$.

Denote $\mathbf{Y} = [y_1^T \ y_2^T]^T = [e^T \ \dot{e}^T]^T$, we have:

$$\begin{cases} \dot{y}_1 = y_2 \\ \dot{y}_2 = -K_p y_1 - K_v y_2 + (\ddot{e} + K_v \dot{e} + K_p e) \end{cases} \quad (30)$$

The system state space equations can be written as:

$$\dot{\mathbf{Y}} = \mathbf{A}\mathbf{Y} + \mathbf{B}\hat{M}_X^{-1}(\tilde{W}^T \Phi + \Delta\delta - \mu) \quad (31)$$

where $\mathbf{A} = \begin{bmatrix} \mathbf{0} & \mathbf{I}_{6 \times 6} \\ -\mathbf{K}_p & -\mathbf{K}_v \end{bmatrix}$, $\mathbf{B} = \begin{bmatrix} \mathbf{0}_{6 \times 6} \\ \mathbf{I}_{6 \times 6} \end{bmatrix}$, The

approximation errors of the system can be regard as external disturbance, and $\Xi = \hat{M}_X^{-1} \Delta\delta \in L_2[0, \infty)$, there

is a positive constant \mathcal{O}_δ let $\int_0^\infty \|\hat{M}_X^{-1} \Delta\delta\|^2 dt \leq \mathcal{O}_\delta$.

Robust item is designed as follow:

$$\mu = \hat{M}_X \mathbf{R}^{-1} \mathbf{B}^T \mathbf{P} \mathbf{y} \quad (32)$$

$\mathbf{R} = \mathbf{R}^T$ is gain matrix, positive definite matrix \mathbf{P} satisfy the Riccati equation:

$$\mathbf{P}\mathbf{A} + \mathbf{A}^T \mathbf{P} + \mathbf{P}\mathbf{B} \left(\frac{1}{\omega^2} \mathbf{I}_{3 \times 3} - 2\mathbf{R}^{-1} \right) \mathbf{B}^T \mathbf{P} = -\mathbf{Q} \quad (33)$$

where ω is disturbance attenuation index \mathbf{Q} is positive definite matrix

Choosing the weights of the adaptive law as follow:

$$\dot{\mathbf{W}} = \mathbf{A}^{-1} \Phi \mathbf{y}^T \mathbf{P} \mathbf{B} \hat{M}_X^{-1} \quad (34)$$

To verify the stability and H_∞ tracking characteristics of the system, the Lyapunov function is defined as follow:

$$V = \frac{1}{2} \mathbf{y}^T \mathbf{P} \mathbf{y} + \frac{1}{2} \text{tr}(\tilde{W}^T \mathbf{A} \tilde{W}) \quad (35)$$

Differentiating Eq.(35)

$$\begin{aligned} \dot{V} &= \frac{1}{2} \dot{\mathbf{y}}^T \mathbf{P} \mathbf{y} + \frac{1}{2} \mathbf{y}^T \dot{\mathbf{P}} \mathbf{y} + \frac{1}{2} \text{tr}(\tilde{W}^T \mathbf{A} \dot{\tilde{W}}) \\ &= \frac{1}{2} \left[\mathbf{A} \mathbf{y} + \mathbf{B} \hat{M}_X^{-1} (\tilde{W}^T \Phi + \Delta\delta - \mu) \right]^T \mathbf{P} \mathbf{y} + \\ &\quad \frac{1}{2} \mathbf{y}^T \mathbf{P} \left[\mathbf{A} \mathbf{y} + \mathbf{B} \hat{M}_X^{-1} (\tilde{W}^T \Phi + \Delta\delta - \mu) \right] + \text{tr}(\tilde{W}^T \mathbf{A} \dot{\tilde{W}}) \\ &= \frac{1}{2} \mathbf{y}^T \left[\mathbf{P}\mathbf{A} + \mathbf{A}^T \mathbf{P} + \mathbf{P}\mathbf{B} \left(\frac{1}{\omega^2} \mathbf{I}_{3 \times 3} - 2\mathbf{R}^{-1} \right) \mathbf{B}^T \mathbf{P} \right] \mathbf{y} - \\ &\quad \frac{1}{2} \left(\frac{1}{\omega} \mathbf{B}^T \mathbf{P} \mathbf{y} - \omega \Xi \right)^T \left(\frac{1}{\omega} \mathbf{B}^T \mathbf{P} \mathbf{y} - \omega \Xi \right) + \frac{1}{2} \omega^2 \Xi^T \Xi \\ &\leq -\frac{1}{2} \mathbf{y}^T \mathbf{Q} \mathbf{y} + \frac{1}{2} \omega^2 \Xi^T \Xi \end{aligned} \quad (36)$$

Integrating Eq.(36) from 0 to T :

$$V(T) - V(0) \leq -\frac{1}{2} \int_0^T \mathbf{y}^T \mathbf{Q} \mathbf{y} + \frac{1}{2} \omega^2 \int_0^T \Xi^T \Xi dt \quad (37)$$

Cause $V(T) \geq 0$, we have:

$$\frac{1}{2} \int_0^T \mathbf{y}^T \mathbf{Q} \mathbf{y} \leq V(0) + \frac{1}{2} \omega^2 \int_0^T \Xi^T \Xi dt$$

So the system tracking errors satisfy H_∞ tracking characteristics of the system.

$\bar{\tau}$ and τ can be rewritten as:

$$\bar{\tau} = [\bar{\tau}_a^T \ \bar{\tau}_b^T]^T \quad (38)$$

$$\tau = [\tau_o^T \ \tau_L^T \ \tau_R^T]^T \quad (39)$$

From the preview derivation process, we have:

$$\bar{\tau}_o = \bar{\tau}_a \quad (40)$$

$$\begin{bmatrix} \tau_L \\ \tau_R \end{bmatrix} = \mathbf{E} \mathbf{Q}^T (\mathbf{Q} \mathbf{E} \mathbf{Q}^T)^{-1} \bar{\tau}_b \quad (41)$$

Where, \mathbf{E} is gain matrix, $\mathbf{Q} = [\mathbf{I}_{3 \times 3} \ (\mathbf{N}_R^{-1} \mathbf{N}_L)^T]$.

4 NUMERICAL SIMULATION

A planar free-floating dual-arm space robot system is experimentally simulated by using the proposed controller. The physical parameters of . The physical parameters of closed-chain system are defined as $d_0 = 1.062\text{m}$, $d_L = 0.5\text{m}$, $d_R = 0.5\text{m}$, $l_i = 2\text{m}$ ($i = 1, 2, 4, 5$), $l_j = 0.5\text{m}$ ($j = 3, 6$), $m_0 = 200\text{kg}$, $m_m = 50\text{kg}$, $m_i = 20\text{kg}$ ($i = 1, 2, 4, 5$), $m_j = 5\text{kg}$ ($j = 3, 6$), $I_0 = 50\text{kg} \cdot \text{m}^2$, $I_m = 10\text{kg} \cdot \text{m}^2$, $I_i = 10\text{kg} \cdot \text{m}^2$ ($i = 1, 2, 4, 5$), $I_j = 2\text{kg} \cdot \text{m}^2$ ($j = 3, 6$).

The initial state of base and load is:

$$X = [0.3\text{m} \quad 0.3\text{m} \quad 10^\circ \quad 0.5224\text{m} \quad 3.5753\text{m} \quad -10^\circ]^T$$

The desired state of base and load is:

$$X_d = [0\text{m} \quad 0\text{m} \quad 0^\circ \quad 0\text{m} \quad 3\text{m} \quad 0^\circ]^T$$

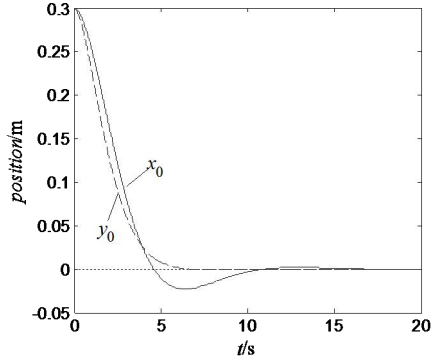


Figure4: Position tracking of base

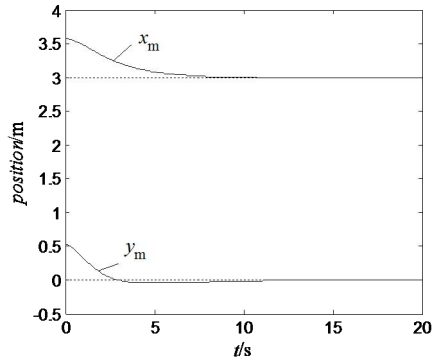


Figure5: Position tracking of load

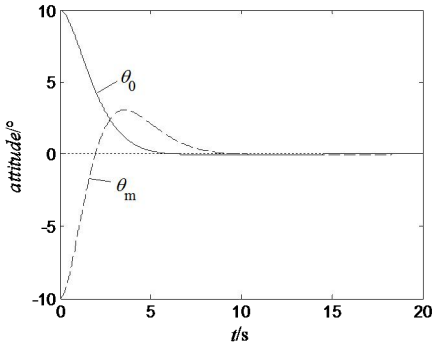


Figure6: Attitude tracking of base and load

Fig.4-6 show the position and attitude of load and base reach the specified location and be stable. These mean that use the proposed control method, the dual-arm space robot could finish the task precisely.

5 CONCLUSION

In this paper, the dynamics and kinematics of dual-arm space robot with closed-chain during transfer a target process is analyzed by closed-loop constraint

equations and the Lagrange equation. The proposed control method can be extended to multi-arm and multi-link space robot system. The control scheme dispense with accurate system model or linear parameterization of the system dynamic equations.

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