Comparative Study of Adaptive Consensus Control of Euler-Lagrange Systems on Directed Network Graph

Yoshihiko Miyasato

The Institute of Statistical Mathematics, Tachikawa, Tokyo, 190-8562, Japan

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ABSTRACT

A comparative study between adaptive consensus control of multi-agent systems composed of fully actuated mobile robots which are described as a class of Euler-Lagrange systems on directed network graphs based on the notion of inverse optimal $H_{\infty}$ control criterion (Controller I), and the similar control strategy without $H_{\infty}$ control criterion (Controller II), is given in this paper. Controller I is deduced as a solution of certain $H_{\infty}$ control problem, where estimation errors of tuning parameters are considered as external disturbances to the process. Asymptotic properties, stability and robustness to unknown parameters are discussed for Controller I and Controller II.

1 Introduction

Motivated by crowd behaviors of animals, birds and fish, cooperative control problems of multi-agent systems are active research fields, and plenty of control strategies have been proposed in those areas, such as formation control, task assignment, traffic control, and scheduling et al. (for example, [1]-[11]). Among those, distributed consensus tracking of multi-agent systems with restricted communication networks, has been a basic and important issue, and various research results have been developed for various processes and under various conditions such as [1], [12]-[21]. In those works, adaptive control methodologies were also investigated in order to deal with uncertainties of agents, and stability of control systems was assured via Lyapunov function analysis. Furthermore, robustness properties of the control schemes were also examined. However, so much attention does not have been paid on control performance such as optimal property or transient performance in those research works, and especially, the case of multi-agent systems composed of processes with unknown and different system parameters on directed information network graphs, does not have been investigated in detail in the previous works.

The purpose of the paper is to provide a comparative study of two types of adaptive consensus control schemes (Controller I and Controller II) of multi-agent systems composed of fully actuated mobile robots which are described as a class of Euler-Lagrange systems [22] on directed network graphs. Controller I is constructed based on the notion of inverse optimal $H_{\infty}$ control criterion [23], [24], and Controller II is synthesized without $H_{\infty}$ control criterion. Controller I is derived as a solution of certain $H_{\infty}$ control problem, where estimation errors of tuning parameters are considered as external disturbances to the process. Asymptotic properties, stability and robustness to unknown parameters are discussed for those two control strategies (Controller I and Controller II).

The proposal of Controller I is an extension of our previous work [25], [26], in which the first-order or second-order linear or nonlinear regression models on directed network graphs were considered. This manuscript is also an extended version of the conference paper [1] (proposal of Controller I), and especially the present paper adds a comparative study of the two types of adaptive consensus control strategies (proposal and comparison of Controller I and Controller II), and highlights robustness properties of Controller I.

2 Multi-Agent Systems and Network Graphs

Mathematical preliminaries on information network graph of multi-agent systems are summarized [17], [20], [21]. As a model of interaction among agents, a weighted directed graph $\mathcal{G} = (V, \mathcal{E})$ is considered,
where \( V = \{1, \ldots, N\} \) is a node set which corresponds to a set of agents, and \( E \subseteq V \times V \) is an edge set. An edge \((i, j) \in E\) means that agent \( j \) can obtain information from \( i \), but not necessarily vice versa. In the edge \((i, j), i \) is called as a parent node and \( j \) is called as a child node, and the in-degree of the node \( i \) is the number of its parents, and the out-degree of \( i \) is the number of its children. An agent which has no parent (or with the in-degree 0), is denoted as a root. A directed path is a sequence of edges in the form \((i_1,i_2), \ldots, (i_l,i_{l+1}) \in E \) where \( i_j \in V \). The directed graph is called strongly connected, if there is always a directed path between every pair of distinct nodes. A directed tree is a directed graph where every node has exactly one parent except for a unique root, and the root has directed paths to all other node. An directed spanning tree \( G_S = (V_S,E_S) \) of the directed graph \( G = (V,E) \) is a subgraph of \( G \) such that \( G_S \) is a directed tree and \( V_S = V \).

Associated with the edge set \( E \), a weighted adjacency matrix \( A = [a_{ij}] \in \mathbb{R}^{N \times N} \) is introduced, and the entry \( a_{ij} \) of it is defined by

\[
a_{ij} = \begin{cases} 
>0 & : (j,i) \in E, \\
0 & : otherwise.
\end{cases}
\]

For the adjacency matrix \( A = [a_{ij}] \), the Laplacian matrix \( L = [l_{ij}] \in \mathbb{R}^{N \times N} \) is defined such as

\[
l_{ii} = \sum_{j=1}^{N} a_{ij}, \\
l_{ij} = -a_{ij}, \quad (i \neq j).
\]

Laplacian matrix has at least one zero eigenvalue and all nonzero eigenvalues have positive real parts. Especially, the Laplacian matrix has a simple 0 eigenvalue with the associated eigenvector \( 1 = [1 \cdots 1]^T \), and all other eigenvalues have positive real parts, if and only if the corresponding directed graph has a directed spanning tree.

In this manuscript, a consensus control problem of leader-follower type is considered, where \( y_0 \) is a leader which each agent \( i \in V \) should follow (\( i \) is called a follower). Concerned with the leader, \( a_{i0} \) is defined such as

\[
a_{i0} = \begin{cases} 
>0 & : \text{leader’s information is available to follower } i, \\
0 & : otherwise,
\end{cases} \tag{1}
\]

and from \( a_{i0} \) and \( L \), the matrix \( M \in \mathbb{R}^{N \times N} \) is defined by

\[
M = L + \text{diag}(a_{i0} \cdots a_{i0}). \tag{2}
\]

It is shown that \(-M\) is a Hurwitz matrix, if and only if 1. at least one \( a_{i0} \ (1 \leq i \leq N) \) is positive, and 2. the graph \( G \) has a directed spanning tree with the root \( i = 0 \).

Hereafter, we assume that

1. The graph has a directed spanning tree with the root \( i = 0 \).

2. At least one \( a_{i0} \ (1 \leq i \leq N) \) is positive, that is, the information of the leader \( y_0 \) (\( \bar{y}_0, \bar{y}_0 \)), is available at least one follower \( i \).

3. For the leader, \( y_0, \bar{y}_0, \bar{y}_0 \) are uniformly bounded.

In the following, two adjacency matrices \( A = [a_{ij}] \), \( C = [c_{ij}] \in \mathbb{R}^{N \times N} \) are introduced for a directed graph \( G \), and the corresponding matrices are denoted as \( L_a, C_a \) (Laplacian matrices), and \( M_a, C_a \), respectively.

### 3 Problem Statement

A multi-agent system composed of \( N \) fully actuated mobile robots which are described as a class of Euler-Lagrange systems \([9],[10]\), are considered such that

\[
M_i(y_i)\ddot{y}_i + C_i(y_i,\dot{y}_i)\dot{y}_i = \tau_i, \quad (i = 1, \ldots, N), \tag{3}
\]

where \( y_i \in \mathbb{R}^n \) is an output (a generalized coordinate), \( \tau_i \in \mathbb{R}^n \) is a control input (a force vector), \( M_i(y_i) \in \mathbb{R}^{n \times n} \) is an inertia matrix, and \( C_i(y_i,\dot{y}_i) \in \mathbb{R}^{n \times n} \) is a matrix of Coriolis and centripetal forces. Each component has the next properties as a Euler-Lagrange system.

**Properties of Euler-Lagrange Systems** \([22]\)

1. \( M_i(y_i) \) is a bounded, positive definite, and symmetric matrix.
2. \( M_i(y_i) - 2C_i(y_i,\dot{y}_i) \) is a skew symmetric matrix.
3. The left-hand side of \((3)\) can be written into

\[
M_i(y_i)a_i + C_i(y_i,\dot{y}_i)b_i = Y_i(y_i,\dot{y}_i,a_i,b_i)\theta_i, \tag{4}
\]

where \( Y_i(y_i,\dot{y}_i,a_i,b_i) \) is a known function of \( y_i, \dot{y}_i, a_i, b_i \) (a regressor matrix), and \( \theta_i \) is an unknown system parameter vector.

The control objective is to achieve consensus tracking of the leader-follower type for the unknown multi-agent system (unknown \( \theta_i \)) such as

\[
y_j \to y_i, \quad (i, j = 1, \ldots, N), \tag{5}
\]

\[
y_i \to \bar{y}_0, \quad (i = 1, \ldots, N), \tag{6}
\]

under the limited communication structure \( G \) among agents.

**Remark 1** More generalized Euler-Lagrange systems including damping terms and gravitational forces, can be also considered in the present framework. However, for simplicity of notations, the description \((3)\) is to be adopted hereafter, since the generalized Euler-Lagrange systems are written in the similar form to \((4)\).
4 Control Law and Error Equation

As the first step of the design of Controller I and Controller II, an estimator of $\hat{y}_0$ (the leader’s information) is developed via available data from the follower $i$. A similar estimation procedure was presented in [16].

$$\hat{z}_i(t) = -\beta \sum_{j=i}^N c_{ij} (\hat{z}_j(t) - \hat{z}_i(t))$$

$$-\beta c_{i0} (\hat{z}_i(t) - \hat{y}_0(t)) + n_{i0} \hat{y}_0(t),$$

(7)

where $\hat{z}_i$ is a current estimate of $\hat{y}_0$, and is synthesized from the data available to the follower $i$. $c_{ij}$ ($1 \leq i \leq N, 0 \leq j \leq N$) is defined as the entry of the adjacency matrix $C$ deduced from the directed graph $G$, and $\beta > 0$ is a design parameter. Associated with $c_{i0}$, $n_{i0}$ is defined such as

$$n_{i0} = \begin{cases} 1 & : c_{i0} > 0, \\ 0 & : \text{otherwise}. \end{cases}$$

(8)

By employing the estimate $\hat{z}_i$, the control scheme is constructed as follows:

$$\hat{y}_r(t) = \hat{z}_i(t) - \alpha \sum_{j=0}^N a_{ij} (\hat{y}_i(t) - y_i(t)),$$

(9)

$$s_i(t) = \hat{y}_i(t) - \hat{y}_r(t),$$

(10)

$$\tau_i(t) = Y_i(t) \hat{\theta}(t) + v_i(t),$$

(11)

$$Y_i(t) = Y_i(\hat{y}_i, \hat{y}_r, v_i),$$

(12)

where $a_{ij}$ ($1 \leq i \leq N, 0 \leq j \leq N$) is defined similarly from the entry of the adjacency matrix $A$ and $\alpha > 0$ is a design parameter. $\hat{\theta}_i$ is denoted as a current estimate of unknown $\theta_i$, and $v_i$ is a stabilizing signal which is to be determined later based on the notion of inverse optimal criterion (Controller I), or without $H_\infty$ criterion (Controller II). An estimation error between the leader $\hat{y}_0$ and the estimate $\hat{z}_i$ is defined by

$$\bar{z}_i(t) = \hat{z}_i(t) - \hat{y}_0(t),$$

(13)

and the following relations are deduced for $s_i$ and $\bar{z}_i$.

$$\bar{z}_i(t) = -\beta \sum_{j=1}^N c_{ij} \bar{z}_j(t) - \bar{z}_i(t)$$

$$-\beta c_{i0} \bar{z}_i(t) + (n_{i0} - 1) \hat{y}_0(t),$$

(14)

$$M_i y_i \bar{s}_i(t) + C_i \hat{y}_i(t) s_i(t)$$

$$= v_i(t) + Y_i(t) \bar{\theta}(t) - \bar{\theta}_i(t).$$

(15)

Then, the overall representations of the multi-agent system are given as follows:

$$\dot{\bar{z}}(t) = -\beta (M_i \otimes I) \bar{z}(t) + [(N_0 - I) \otimes I] \hat{y}_0(t),$$

(16)

$$Ms(t) + Cs(t) = Y_i(t) \bar{\theta} - \theta(t) + v_i(t),$$

(17)

where

$$\ddot{z} = [\bar{z}_1^T, \ldots, \bar{z}_N^T]^T,$$

$$s = [s_1^T, \ldots, s_N^T]^T,$$

$$M = \text{block diag}(M_1, \ldots, M_N), \quad (M_i \equiv M_i(y_i)),$$

$$C = \text{block diag}(C_1, \ldots, C_N), \quad (C_i \equiv C_i(y_i, \hat{y}_i)),$$

$$Y = \text{block diag}(Y_1, \ldots, Y_N),$$

$$\theta = \text{block diag}(\theta_1^T, \ldots, \theta_N^T)^T,$$

$$N_0 = [n_{01}, \ldots, n_{0N}]^T,$$

$$1 = [1, \ldots, 1]^T,$$

$$v = [v_1^T, \ldots, v_N^T]^T,$$

(18)-(26)

and $\otimes$ denotes Kronecker product.

5 Adaptive $H_\infty$ Consensus Control for Euler-Lagrange Systems (Controller I)

In this section a design scheme of adaptive consensus control based on inverse optimal $H_\infty$ criterion (Controller I) is provided. Stability analysis of the overall control system and deduction of Controller I are composed of four steps. As the first step, for stability analysis of $s$ and the related terms, a positive function $W_0$ is defined such as

$$W_0(t) = \frac{1}{2} s(t)^T Ms(t).$$

(27)

Then, the time derivative of $W_0$ along its trajectory is derived as follows:

$$W_0(t) = s(t)^T Y(t) [\bar{\theta} - \theta(t)] + v(t).$$

(28)

By considering the evaluation of $W_0$ [28], the next virtual system is introduced.

$$\dot{s} = f + g_1 d + g_2 v,$$

(29)

$$f = 0,$$

(30)

$$g_1 = Y, \quad g_2 = I,$$

(31)

$$d = (\bar{\theta} - \theta).$$

(32)

The virtual system is to be stabilized via a control input $v$ based on $H_\infty$ criterion, where $d$ is considered as an external disturbance to the process. For that purpose, the following Hamilton-Jacobi-Isaacs (HJI) equation and its solution $V_0$ are introduced.

$$\mathcal{L}_f V_0 + \frac{1}{2} \left( \frac{\|L_{g_1} V_0\|^2}{\gamma^2} - (\mathcal{L}_{g_2} V_0) R_1^{-1} (\mathcal{L}_{g_2} V_0)^T \right) + q = 0,$$

(33)

$$V_0 = \frac{1}{2} s^T s,$$

(34)

where $q$ and $R$ are a positive function and a positive definite matrix respectively, and those are deduced from HJI equation based on the notion of inverse optimality for the given solution $V_0$ and a positive constant $\gamma$. The substitution of the solution $V_0$ [34] into HJI equation [33] yields

$$\frac{1}{4} s^T \left( \frac{Y^T}{\gamma^2} - R^{-1} \right) s + q = 0.$$

(35)

From [35], $R$ and $q$ are obtained such as

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where \( K \) is a diagonal positive definite matrix (a design parameter). From \( R, v \) is derived as a solution of the corresponding \( H_{\infty} \) control problem as follows (Controller I):

\[
\begin{align*}
R &= \left( \frac{Y Y^T}{\gamma^2} + K \right)^{-1}, \\
q &= \frac{1}{4} s^T K s,
\end{align*}
\]  

(36)

(37)

Then, via HJI equation, the time derivative of \( W_0 \) is evaluated as follows:

\[
W_0 = -q - v^T R v + \left( v + \frac{1}{2} R^{-1} s \right) R \left( v + \frac{1}{2} R^{-1} s \right)^T + \gamma^2 \|d\|^2 - \gamma^2 \left\| d - \frac{Y T s}{2 \gamma^2} \right\|^2,
\]  

(38)

and it follows that \( s \) is bounded for bounded \( \theta \) and for the stabilizing signal \( v \).

Next, for stability analysis of the estimation error \( \dot{\theta} \), a positive function \( V_1 \) is introduced such as

\[
V_1 = \bar{\theta}^T (P_c \otimes I) \bar{\theta},
\]  

(40)

\[
P_c M_c + M_c^T P_c = I, \quad (P_c = P_c^T > 0).
\]  

(41)

There exists a positive definite and symmetric matrix \( P_c \) satisfying \([11] \), since \(-M_c\) is Hurwitz. Then, the time derivative of \( V_1 \) along its trajectory is evaluated as follows:

\[
\begin{align*}
\dot{V}_1 &= -\beta \| \bar{\theta} \|^2 - 2 \bar{\theta}^T (P_c \otimes I) ((N_0 - 1) \otimes I) \bar{\theta} - \frac{1}{2} \| \bar{\theta} \|^2 + 2 \|P_c \otimes I\|^2 \|((N_0 - 1) \otimes I) \bar{\theta} \|^2,
\end{align*}
\]  

(42)

and it is shown that \( \dot{\theta} \) is bounded for bounded \( \theta \).

Thirdly, for stability analysis of the control error \( y_i - \hat{y}_0 \) and the related terms, \( \bar{\eta}, \bar{\theta} \) are defined by

\[
\begin{align*}
\bar{\eta} &= y_i - \hat{y}_0, \\
\bar{\theta} &= \begin{bmatrix} \bar{\eta}_1^T, \cdots, \bar{\eta}_N^T \end{bmatrix}^T.
\end{align*}
\]  

(43)

(44)

Then, the following relation holds

\[
\begin{align*}
\dot{\bar{\eta}} &= \gamma + 2 - \alpha (M_a \otimes I) \bar{\eta},
\end{align*}
\]  

(45)

and \(-M_a\) is shown to be Hurwitz because of the assumption of the network graph \( G \). From that property, a positive function \( V_2 \) is defined by

\[
\begin{align*}
V_2 &= \gamma^T (P_a \otimes I) \bar{\eta}, \\
P_a M_a + M_a^T P_a = I, \quad (P_a = P_a^T > 0).
\end{align*}
\]  

(46)

(47)

Similarly to the previous case \((M_a)\), there exists a positive definite and symmetric matrix \( P_a \) satisfying \([47] \), since \(-M_a\) is Hurwitz. Then, the time derivative of \( V_2 \) along its trajectory is evaluated as follows:

\[
\begin{align*}
\dot{V}_2 &= -\alpha \| \bar{\eta} \|^2 + 2 \bar{\eta}^T (P_a \otimes I) (s + \hat{\theta}) - \frac{1}{2} \| \bar{\eta} \|^2 + \frac{4}{\alpha} \|P_a \otimes I\|^2 (\|s\|^2 + \|\hat{\theta}\|^2).
\end{align*}
\]  

(48)

From the three stages of stability analysis (the evaluations of \( W_0, V_1, V_2 \)), the next theorem is obtained.

**Theorem 1** The nonlinear control system composed of the control laws \([7, 9, 10, 11, 12, 38] \) (Controller I) is uniformly bounded for an arbitrary bounded design parameter \( \theta \), and bounded \( \gamma_0, \gamma_0, \gamma_0 \), and \( v \) is an optimal control input which minimizes the following cost functional \( J \).

\[
J(t) = \sup_{d_1, d_2, \in} \left[ t \int_0^t \|q + v^T R v\| d\tau + W_0(t) \right]
\]  

(49)

Also the next inequality holds.

\[
\int_0^t \|q + v^T R v\| d\tau + W_0(t) \leq \gamma^2 \int_0^t \|d\|^2 d\tau + W_0(0).
\]  

(50)

Theorem 1 denotes the properties of the proposed nonlinear control system \([7, 9, 10, 11, 12, 38] \) (Controller I), where the tunings of \( \theta \) is not included (or not necessarily required).

Next, the tuning law of \( \theta \) is determined as follows:

\[
\dot{\theta}(t) = \text{Pr} \left( -\Gamma Y (T^T s)(t) \right),
\]  

(51)

where \( \text{Pr}(\cdot) \) is a projection operation in which the tuning parameter \( \theta \) is constrained to a bounded region deduced from upper-bounds of \( \|\theta\| \). As the fourth step of stability analysis of the overall control system, a positive function \( W_1 \) is defined by

\[
W_1(t) = \frac{1}{2} s(t)^T s(t) + \frac{1}{2} \left\{ \dot{\theta}(t) - \theta \right\}^T \Gamma^{-1} \left\{ \dot{\theta}(t) - \theta \right\},
\]  

(52)

and the time derivative of \( W_1 \) along its trajectory is evaluated such as

\[
\dot{W}_1(t) \leq -\frac{1}{2} s(t)^T R^{-1} s(t) \leq 0.
\]  

(53)

From the four stages of stability analysis (the evaluations of \( W_0, W_1, V_1, V_2 \)), the next theorem is obtained.

**Theorem 2** The adaptive control system composed of the control laws \([7, 9, 10, 11, 12, 38] \), and the tuning law of \( \theta \) \([51] \) (Controller I), is uniformly bounded for bounded \( \gamma_0, \gamma_0, \gamma_0 \), and it follows that

\[
\lim_{t \rightarrow \infty} s(t) = 0.
\]  

(54)

Especially, if \( \gamma_0(t) = 0 \) or the information of the leader \( \gamma_0 \) is available for all followers \(((N_0 - 1) \otimes I) \gamma_0 = 0 \), then it follows that

\[
\lim_{t \rightarrow \infty} \dot{\gamma}(t) = \lim_{t \rightarrow \infty} \bar{\eta}(t) = 0.
\]  

(55)
and the asymptotic consensus tracking is achieved. Otherwise, when \( \tilde{y}_0(t) \neq 0 \) and the information of \( \tilde{y}_0 \) is not available for all followers \((N_0 - 1) \otimes I \) \( \tilde{y}_0 \neq 0 \), then the next relation holds.

\[
\|y\| \sim \text{const.} \frac{1}{\alpha \beta} \|((N_0 - 1) \otimes I)\tilde{y}_0\|,
\]

\[
\|\dot{y}\| \sim \text{const.} \frac{1}{\beta} \|((N_0 - 1) \otimes I)\tilde{y}_0\|.
\]

Theorem 2 denotes the properties of the proposed adaptive control system \((7), (9), (10), (11), (12), (38), (51)\) (Controller I), and states that the asymptotic consensus tracking is achieved under the specified condition \(((N_0 - 1) \otimes I)\tilde{y}_0 = 0\), and also shows that the approximate consensus tracking with the ratios of \(1/(\alpha \beta), 1/\beta > 0\) is still assured, even if that condition is not satisfied \(((N_0 - 1) \otimes I)\tilde{y}_0 \neq 0\).

Remark 2 It should be noted the proposed control scheme and the adaptation scheme (Controller I) are all implemented in a distributed fashion, where availabilities of signals for each agent \( i \) are highly restricted and prescribed by the directed graph \( G \).

Remark 3 The objective of the design of Controller I, is to obtain an adaptive control structure whose stability is not seriously affected by the estimation errors of the tuning parameters. For that purpose, in order to attenuate the effects of the estimation errors, the control scheme is to be deduced as a solution of certain \( H_{\infty} \) control problem, where \( L_2 \) gains from the estimation errors of tuning parameters to the generalized output is prescribed by a positive constant \( \gamma \) (design parameter).

Remark 4 Of course, \( f \) in Theorem 1 is a fictitious cost functional, since \( d \) is not an actual disturbance but an estimation error of the tuning parameter, and since it is not generally included in \( L^2[0, \infty) \). Nevertheless, \( v \), which is derived as a solution for that fictitious \( H_{\infty} \) control problem (Controller I), attain the inequality in Theorem 1, stabilize the total system, and it means that the \( L^2 \) gain from the disturbance \( d \) to the generalized output \( \sqrt{q^2 + \dot{v}^T R \dot{v}} \) is prescribed by the positive constant \( \gamma \).

6 Adaptive Consensus Control for Euler-Lagrange Systems without \( H_{\infty} \) Criterion (Controller II)

The adaptive consensus control strategy without \( H_{\infty} \) control criterion (Controller II) is easily deduced by setting \( v \to \infty \) in the synthesis of \( v \) \((38)\) such that

\[
v = -\frac{1}{2} K_s.
\]

In this case, the tuning scheme of \( \hat{\theta} \) \((51)\) is required to assure uniform boundedness of the total control system. The time derivative of \( W_1 \) \((52)\) with the tuning law \( (51) \) gives

\[
W_1(t) \leq -\frac{1}{2} s(t)^T K_s \leq 0.
\]

Then, by applying the same stability analysis via the evaluations of \( W_1, V_1, V_2 \), the following theorem is obtained.

Theorem 3 The adaptive control system composed of the control laws \((7), (9), (10), (11), (12), (38)\), and the tuning law of \( \hat{\theta} \) \((51)\) (Controller II), is uniformly bounded for bounded \( \tilde{y}_0, \tilde{y}_0, y_0 \) and it follows that

\[
\lim_{t \to \infty} s(t) = 0.
\]

Especially, if \( \tilde{y}_0(0) = 0 \) or the information of the leader \( \tilde{y}_0 \) is available for all followers \(((N_0 - 1) \otimes I)\tilde{y}_0 = 0 \), then it follows that

\[
\lim_{t \to \infty} \dot{y}(t) = \lim_{t \to \infty} \tilde{y}(t) = 0,
\]

and the asymptotic consensus tracking is achieved. Otherwise, when \( \tilde{y}_0(0) \neq 0 \) and the information of \( \tilde{y}_0 \) is not available for all followers \(((N_0 - 1) \otimes I)\tilde{y}_0 \neq 0 \), then the next relation holds.

\[
\|y\| \sim \text{const.} \frac{1}{\alpha \beta} \|((N_0 - 1) \otimes I)\tilde{y}_0\|,
\]

\[
\|\dot{y}\| \sim \text{const.} \frac{1}{\beta} \|((N_0 - 1) \otimes I)\tilde{y}_0\|.
\]

Remark 5 Asymptotic properties of the adaptive consensus control without \( H_{\infty} \) criterion \((60), (61), (62), (63)\) (Controller II) are the same as the adaptive \( H_{\infty} \) consensus control version \((54), (55), (56), (57)\) (Controller I). However, it should be noted that uniform boundedness of the overall control system via Controller II is not assured for arbitrary bounded design parameters \( \hat{\theta} \). Instead, the tuning of \( \hat{\theta} \) \((51)\) is necessary to attain the stability of consensus tracking by Controller II. Furthermore, Controller II in Theorem 3 does not contain an optimal property for the cost functional \((49)\), and the inequality \((50)\) does not hold.

7 Numerical Example

In order to show the effectiveness of the two types of adaptive consensus control schemes, numerical experiments for Euler-Lagrange systems are performed. A multi-agent system composed of simple Euler-Lagrange systems is considered as follows:

\[
m_i \ddot{y}_i(t) = \tau_i(t), \quad (i = 1, 2, 3),
\]

\[
(y_1(0) = 1, y_2(0) = 0, y_3(0) = -1),
\]

where \( y_i \in \mathbb{R}, \tau_i \in \mathbb{R}, \) and \( m_i \in \mathbb{R} \) is an unknown system parameter. Associated with the information network structure \((\text{Fig.1})\), the adjacency matrix \( A = [a_{ij}]= C \) and \( a_{i0} = c_{i0} \) are chosen such that

\[
A = \begin{bmatrix}
0 & 1 & 0 \\
1 & 0 & 1 \\
0 & 1 & 0
\end{bmatrix},
\]

\( a_{10} = 1, \ a_{20} = a_{30} = 0. \)
The control objective is to achieve consensus tracking
\[
y_i \to y_j, \quad \dot{y}_i \to \dot{y}_j, \\
y_i \to y_0, \quad \dot{y}_i \to \dot{y}_0,
\]
\((i, j = 1, 2, 3),\)
where the virtual leader \(y_0\) is determined such as
\[
y_0 + 2\dot{y}_0(t) + y_0(t) = \sin t.
\]

The design parameters are chosen as follows:
\[
\Gamma = 10I, \quad K = 5I, \quad \alpha = \beta = 10, \quad y_1 = 1.
\]
As system parameters \(m_i (i = 1 \sim 3)\), we consider both time-invariant and time-varying cases such that
\[
m_1 = 1, \quad m_2 = 2, \quad m_3 = 3, \quad \text{(time-invariant case)},
\]
\[
m_1 = f_m(t), \quad m_2 = 2f_m(t), \quad m_3 = 3f_m(t), \quad \text{(time-varying case)},
\]
where
\[
f_m(t) = \begin{cases} 
2 & 0 \leq t < 2.5, \quad 5 \leq t < 7.5, \cdots, \\
1 & 2.5 < t \leq 5, \quad 7.5 < t \leq 10, \cdots. 
\end{cases}
\]

Numerical simulations were carried out by utilizing MATLAB and Simulink of the MathWorks, Inc., and the solver is Dormand-Prince method with the adaptive step-size integration algorithm (ode45).

The simulation results of Controller I (Theorem 2) are shown in Fig.2 (time-invariant case) and Fig.4 (time-varying case). For comparison, the cases of Controller II (the adaptive control systems which do not contain \(H_\infty\) control scheme), are shown for both cases; Fig.3 (time-invariant case) and Fig.5 (time-varying case). In those figures (Fig.2 ~ Fig.5), the plots were response curves of \(y_1, y_2, y_3\) and \(y_0\) (vertical axes) versus time \(t\) (horizontal axes).

From those results, it is seen that Controller I (\(H_\infty\), adaptive control strategy) achieves better tracking convergence property together with robustness to abrupt changes of the system parameters compared with the non-\(H_\infty\) control scheme (Controller II), and those are owing to disturbance attenuation properties of Controller I, since the performance of Controller I is not significantly degraded by the estimation errors of the tuning parameters in the settings of both time-invariant and time-varying cases (see also Remark 3).

8 Concluding Remarks

The comparative study of two types of adaptive consensus control (Controller I and Controller II) of multi-agent systems composed of fully actuated mobile robots which are described as a class of Euler-Lagrange systems on directed network has been given in this paper. Controller I is deduced based on the notion of inverse optimal \(H_\infty\) control criterion, and is derived as a solution of certain \(H_\infty\) control problem, where estimation errors of tuning parameters are considered as external disturbances to the process. On the contrary, Controller II is synthesized without \(H_\infty\) control criterion. Asymptotic properties, stability and robustness to unknown parameters are discussed for those two control strategies (Controller I and Controller II). Although asymptotic properties of those two controllers are similar to each others, Controller I achieves better tracking convergence property together with robustness to abrupt changes of the system parameters compared with Controller II, and those are owing to disturbance attenuation properties of Controller I and the problem setting of \(H_\infty\) criterion.
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