# Reliable Consensus Problem for Multi-Agent Systems with Sampled-Data

S. H. Lee, M. J. Park, O. M. Kwon

**Abstract**—In this paper, reliable consensus of multi-agent systems with sampled-data is investigated. By using a suitable Lyapunov-Krasovskii functional and some techniques such as Wirtinger Inequality, Schur Complement and Kronecker Product, the results of such system are obtained by solving a set of Linear Matrix Inequalities (LMIs). One numerical example is included to show the effectiveness of the proposed criteria.

*Keywords*—Multi-agent, Linear Matrix Inequalities (LMIs), Kronecker Product, Sampled-Data, Lyapunov method.

#### I. INTRODUCTION

N recent years, many problem of multi-agent systems has I received considerable attentions due to their extensive applications in cooperative control of mobile autonomous robots, the design of distributed sensor networks, spacecraft formation flying and so on. A main problem in its systems is the consensus problem that it is the agreement of a group of agents on their states of leader by interaction [1]-[5]. Nevertheless, this problem recently has been applied in various fields such as vehicle systems [6], [7], groups of mobile autonomous agent [8], networked control systems [9], other applications. However, it is considered to use the problems of multi-agent systems due to the limited speed of information processing in the implementation of this system. Specially, it is well known that time-delay often causes unwanted signal like oscillations and noises of the system [4]. Thus, it is essential to study them. So motivated by this mentioned above, in this paper, new consensus problem for multi-agent systems with both sampled -data and reliable will be studied.

At first, in industrial process control, the digital control, digital filtering, and signal processing are widely used, which makes the closed-loop systems hybrid so-called sampled-data system; its states suffer successive impulses at fixed times. The sampled-data system is a hybrid one involving continuous time and discrete time signals [10].

Next, networked control systems use data networks to close both information and control loops. Networked control systems integrate information, communications and control with control loops being closed through the network [11]. They are becoming increasingly important in industrial process control because of their cost-effectiveness, reduced weight and power requirement, simple installation and maintenance and high reliability. The problem of designing reliable control systems has been attracted since practical systems often have actuator failures [11], [12]. It has been known that the class of reliable control systems is to stabilize the systems against actuator failures or to design fault-tolerant control systems. So the actuator failure model which consists of a scaling factor with upper and lower bounds to the signal to be measured or to the control action is introduced [13].

In this paper, reliable consensus of multi-agent systems with sampled-data was supposed. Also, in order to better results, this paper was used to Wirtinger-based integral inequality.

Notation:  $\mathfrak{R}^n$  is the *n*-dimensional Euclidean space, and  $\mathfrak{R}^{m\times n}$  denotes the set of all  $m \times n$  real matrices. For symmetric matrices *X* and *Y*, *X* > *Y* means that the matrix X - Y is positive definite.  $X^T$  denotes the transpose for *X*. If the context allows it, the dimensions of these matrices are often omitted.  $I_n$ ,  $0_n$  respectively denote  $n \times n$  identity matrix and zero matrix.  $X^{\perp}$  denotes a basis for the null-space of *X*. For a given matrix  $X \in \mathfrak{R}^{n\times n}$ , we define  $X^{\perp} \in \mathfrak{R}^{n\times (n-r)}$  as the right orthogonal complement of *X* by  $XX^{\perp} = 0$ .  $dia \{\cdots\}$  denotes the block diagonal matrix. \* represents the elements the main diagonal of a symmetric matrix.  $\otimes$  denotes the notation of Kronecker product.

### **II. PROBLEM STATEMENTS**

Consider the multi-agent systems with the following dynamic of agent i

$$\dot{x}_i(t) = Ax_i(t) + Bu_i(t), \ i = 1, ..., N,$$
 (1)

where *N* is the number of agents,  $x_i(t) \in \mathbf{R}^n$  is the state of agent *i*,  $u_i(t) \in \mathbf{R}^m$  is the consensus protocol, and  $A \in \mathbf{R}^{n \times n}$  and  $B \in \mathbf{R}^{n \times m}$  are known constant matrices.

An algorithm of consensus protocol can be described as

$$u_i(t) = -\sum_{j \in N_i} g_{ij}(x_i(t) - x_j(t))), \ i = 1, ..., N,$$
(2)

where  $g_{ii}$  are the interconnection weights defining

$$\begin{cases} g_{ij} > 0, \text{ if agent } i \text{ is connected to agent } j, \\ g_{ii} = 0, \text{ otherwise.} \end{cases}$$

The multi-agent system is said to achieve consensus if the following definition.

**Definition 1.** [17], [18] Given an undirected communication graph G, the multi-agent systems (1) are said to be consensus-

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able under the protocol (2) if for any finite  $x_i(0)$ , i = 1,...,N, the control protocol can asymptotically drive all agents close to each other, i.e.,

$$\lim_{t \to \infty} ||x_i(t) - x_j(t)|| = 0, \ i = 1, \dots, N.$$

In this paper, it is concerned that actuator has behavior of faulty. The control input of actuator fault can be described as

$$u^{F}(t) = Ru(t) \tag{5}$$

where R is the actuator fault matrix with

$$R = diag\left\{r_1, r_2, \cdots, r_m\right\}, \ 0 \le \underline{r}_i \le r_i \le \overline{r}_i, \ \overline{r}_i \ge 1, \ \left(i = 1, 2, \cdots, m\right)$$
(6)

where  $\underline{r}_i$  and  $\overline{r}_i$   $(i=1,2,\dots,m)$  are given constants. When  $r_i = 1$ , it means the complete failure of i-th actuator. If  $r_i = 1$ , then i-th actuator is normal.

Let us define

$$R_0 = diag\{r_{10}, r_{20}, \cdots, r_{m0}\}, \ r_{i0} = \frac{\overline{r_i} + r_i}{2}$$
(7)

$$R_{\rm l} = diag\{r_{11}, r_{21}, \cdots, r_{m1}\}, \ r_{i1} = \frac{\overline{r_i} - r_i}{2}$$
(8)

Then, the actuator fault matrix R can be rewritten as

$$R = R_0 + R_1 \vartriangle J \tag{9}$$

where  $\Delta J = diag\{j_1, j_2, \dots, j_m\}, -1 \le j_i \le 1$ .

The updating instant time of the Zero-Order Hold is denoted by  $t_k$ . We assume that the sampling intervals are bounded  $t_{k+1} - t_k \le h_k$ . The state-feedback controller has a form  $u(t) = x(t_k)$ . Defining  $h(t) = t - t_k$ , we have u(t) = x(t - h(t)),  $t_k \le t \le t_{k+1}$ ,  $k = 0, 1, 2, \cdots$ .

We obtain reliable following the consensus of multi-agent systems with sampled-data. With the concept introduced at (3)-(9), let us consider reliable consensus of multi-agent systems with sampled-data and actuator failures given by

$$\dot{x}_{i}(t) = Ax_{i}(t) + \sum_{j \in N_{i}} g_{ij}Bx_{j}(t-h(t)),$$

$$i = 1, ..., N, \quad t_{k} \le t \le t_{k+1} \quad , \quad k = 0, 1, 2, \cdots$$
(10)

Then, the system (10) can be rewritten as

$$\dot{x}(t) = (I_N \otimes A)x(t) + (G \otimes B(R_0 + R_1 \Delta J))x(t - h(t))(11)$$

where  $x(t) = \begin{bmatrix} x_1(t) & x_2(t) & \cdots & x_N(t) \end{bmatrix}^T \in \mathfrak{R}^n$ ,  $G^k = \begin{bmatrix} a_{ij}^k \end{bmatrix}_{N \times N}$ .

Before deriving main results, the following lemmas are introduced.

**Lemma 1. (Reciprocally convex combination) [14]:** For a scalar  $\alpha$  in the interval (0,1), a given matrix  $R \in \Re^{n \times n} > 0$ ,

two matrices  $W_1 \in \Re^{n \times m}$  and  $W_2 \in \Re^{n \times m}$ , for all vector  $\zeta \in \Re^m$ , let us define the function  $\theta(\alpha, \mathbf{R})$  given by:

$$\theta(\alpha, \mathbf{R}) = \frac{1}{\alpha} \zeta^{T} W_{1}^{T} R W_{1} \zeta + \frac{1}{1 - \alpha} \zeta^{T} W_{2}^{T} R W_{2} \zeta$$

Then, if there exists a matrix  $X \in \Re^{n \times m}$ , then the following inequality holds

$$\min_{\alpha \in (0,1)} \theta(\alpha, \mathbf{R}) \ge \begin{bmatrix} W_1 \zeta \\ W_2 \zeta \end{bmatrix}^T \begin{bmatrix} \mathbf{R} & \mathbf{X} \\ * & \mathbf{R} \end{bmatrix} \begin{bmatrix} W_1 \zeta \\ W_2 \zeta \end{bmatrix}$$

**Lemma 2. (Wirtinger inequality):** For a given matrix R > 0, the following inequality holds for all continuously differentiable function W in  $[a,b] \rightarrow \Re^n$ .

$$\int_{a}^{b} w^{T}(u) Rw(u) du$$

$$\geq \frac{1}{b-a} \left( w(b) - w(a) \right)^{T} R \left( w(b) - w(a) \right) + \frac{3}{b-a} \Omega^{T} R \Omega$$
where  $\Omega = w(b) + w(a) - \frac{2}{b-a} \int_{a}^{b} w(u) du$ 

**Lemma 3. (Kronecker product)** [15]: Let  $\otimes$  denote the notation of Kronecker product. Then, the following properties of the Kronecker product are easily established: (*i*)( $\alpha A$ ) $\otimes B = A \otimes (\alpha B)$ ,

 $(ii)(A+B) \otimes C = A \otimes C + B \otimes C,$   $(iii)(A \otimes B)(C \otimes D) = (AC) \otimes (BD),$  $(iv)(A \otimes B)^{T} = A^{T} \otimes B^{T}.$ 

**Lemma 4.** [16]: Let E, H and  $_{F(t)}$  be real matrices of appropriate dimensions, and let  $_{F(t)}$  satisfy  $F^{T}(t)F(t) \le I$ . Then, for any scalar  $\varepsilon > 0$ , the following matrix inequality holds:

$$EF(t)H + H^{\mathrm{T}} + F^{\mathrm{T}}(t)E^{\mathrm{T}} \leq \varepsilon H^{\mathrm{T}}H + \varepsilon^{-1}EE^{\mathrm{T}}$$

## III. MAIN RESULT

In this section, we propose new stability and stabilization criteria for system (9). The notations of several matrices are defined as:

$$\begin{split} e_{i}^{T}(t) &= \begin{bmatrix} x^{T}(t) & x^{T}(t-h(t)) & x^{T}(t-h_{M}) & \frac{1}{h(t)} \int_{t-h(t)}^{t} x(s) ds & \frac{1}{h_{M} - h(t)} \int_{t-h_{N}}^{t-h(t)} x(s) ds & \end{bmatrix}; \\ e_{i} &= \begin{bmatrix} 0_{(i-1)n} & I_{n} & 0_{(5-i)n} \end{bmatrix}^{T}, \ (i = 1, 2, \cdots, 5) , \\ \Phi &= \Phi_{1} + \Phi_{2} + \Phi_{3} + \Phi_{4} , \ \Phi_{1} &= e_{1} (I_{N} \otimes P) \psi^{T} + \psi (I_{N} \otimes P) e_{1}^{T} , \\ \Phi_{2} &= e_{1} (I_{N} \otimes Q_{1}) e_{1}^{T} - (1-h_{d}) e_{2} (I_{N} \otimes Q_{1}) e_{2}^{T} , \\ \Phi_{3} &= e_{1} (I_{N} \otimes Q_{2}) e_{1}^{T} - e_{3} (I_{N} \otimes Q_{2}) e_{3}^{T} , \\ \Phi_{4} &= -\begin{bmatrix} e_{1}^{T} - e_{2}^{T} \\ e_{1}^{T} + e_{2}^{T} - 2e_{1}^{T} \\ e_{2}^{T} - e_{3}^{T} \\ e_{2}^{T} - e_{3}^{T} \\ e_{2}^{T} + e_{3}^{T} - 2e_{6}^{T} \end{bmatrix}^{T} \begin{bmatrix} I_{N} \otimes R_{1} & I_{N} \otimes M \\ * & I_{N} \otimes R_{1} \end{bmatrix} \begin{bmatrix} e_{1}^{T} - e_{2}^{T} \\ e_{1}^{T} + e_{2}^{T} - 2e_{5}^{T} \\ e_{2}^{T} - e_{3}^{T} \\ e_{2}^{T} + e_{3}^{T} - 2e_{6}^{T} \end{bmatrix} , \end{split}$$

$$R_{1} = \begin{bmatrix} I_{N} \otimes R & 0 \\ * & I_{N} \otimes 3R \end{bmatrix}, \quad \Psi = \Psi_{1} + \Psi_{2} ,$$

$$\Psi_{1} = (I_{N} \otimes A)e_{1}^{T} + (G \otimes BR_{0})e_{2}^{T} , \quad \Psi_{2} = (G \otimes BR_{1}\Delta I)e_{2}^{T} ,$$

$$H_{1} = G \otimes BR_{1} , \quad H_{1} = G \otimes BR_{1} , \quad \Psi = \varepsilon_{2} (GG^{T} \otimes BR_{1} (BR_{1})^{T}) ,$$

$$\Xi = \Phi + \varepsilon_{1}e_{1} (G \otimes PBR_{1})^{T} (G \otimes PBR_{1})e_{1}^{T} ,$$

$$\Pi = \begin{bmatrix} \Xi & h_{M}\Psi_{1}^{T} & e_{2}^{T} & e_{2}^{T} \\ h_{M}\Psi_{1} & -(I_{N} \otimes \alpha X) + \Psi & 0 & 0 \\ e_{2} & 0 & -\varepsilon_{1}I & 0 \\ e_{2} & 0 & 0 & -\varepsilon_{2}I \end{bmatrix}$$

Now we have Theorem 1.

**Theorem 1.** For given scalars  $h_M$ ,  $\overline{r_i}$ ,  $\underline{r_i}$ ,  $\alpha$ , and the matrices A, B, G, the agent in the system (11) converge to the state of leader, if there exist positive definite matrices  $P \in \Re^{n \times n}$ ,  $Q_1 \in \Re^{n \times n}$ ,  $Q_2 \in \Re^{n \times n}$ ,  $X \in \Re^{n \times n}$  and positive scalar  $\varepsilon_1$ ,  $\varepsilon_2$ , any matrix  $M \in \Re^{2n \times 2n}$ . Then system is asymptotically stable for ~~~when satisfying the following LMIs:

$$\Pi < 0 \tag{12}$$

$$\begin{bmatrix} I_N \otimes R_1 & I_N \otimes M \\ * & I_N \otimes R_1 \end{bmatrix} \ge 0$$
(13)

**Proof:** Let us consider the following Lyapunov-krasovskii functional candidate as

$$V(x(t)) = V_1(x(t)) + V_2(x(t)) + V_3(x(t)) + V_4(x(t))$$
(14)

where

$$V_{1}(x(t)) = x^{T}(t)(I_{N} \otimes P)x(t)$$

$$V_{2}(x(t)) = \int_{t-h_{U}}^{t} x^{T}(s)(I_{N} \otimes Q_{1})x(s)ds$$

$$V_{3}(x(t)) = \int_{t-h_{U}}^{t} x^{T}(s)(I_{N} \otimes Q_{2})x(s)ds$$

$$V_{4}(x(t)) = h_{M} \int_{t-h}^{t} \int_{t}^{t} \dot{x}^{T}(u)(I_{N} \otimes R)\dot{x}(u)duds$$

By using Lemma 3 and 4 the time derivative of  $V_1$  is calculated as

$$\dot{V}_{1}(x(t)) = \zeta^{T}(t)(e_{1}(I_{N} \otimes P)\psi^{T} + \psi(I_{N} \otimes P)e_{1}^{T})\zeta(t)$$

$$= \zeta^{T}(t)(\Phi_{1} + e_{1}(I_{N} \otimes P)\psi_{2}^{T} + \psi_{2}(I_{N} \otimes P)e_{1}^{T})\zeta(t)$$

$$(15)$$

By using Lemma 4, the upper-bound of time derivative of  $V_2$  is calculated as

$$\dot{V}_{2}(x(t)) = \frac{d}{dt} \int_{t-h(t)}^{t} x^{T}(s) (I_{N} \otimes Q_{1}) x(s) ds$$

$$\leq \zeta^{T}(t) (e_{1}(I_{N} \otimes Q_{1}) e_{1}^{T} - (1-h_{d}) e_{2}(I_{N} \otimes Q_{1}) e_{2}^{T}) \zeta(t)$$

$$= \zeta^{T}(t) \Phi_{2}\zeta(t)$$
(16)

By using Lemma 4, the upper-bound of time derivative of  $V_3$  is calculated as

$$\dot{V}_{3}(x(t)) = \zeta^{T}(t) \Big( e_{1} \big( I_{N} \otimes Q_{1} \big) e_{1}^{T} - e_{3} \big( I_{N} \otimes Q_{2} \big) e_{3}^{T} \big) \zeta(t)$$

$$= \zeta^{T}(t) \Phi_{3} \zeta(t)$$
(17)

By using Lemma 4, the time derivative of  $V_4$  is calculated as

$$\dot{V}_4(x(t)) = h_M^2 \dot{x}^T(t) (I_N \otimes R) \dot{x}(t) - h_M \int_{t-h}^t \dot{x}^T(u) (I_N \otimes R) \dot{x}(u) du$$

Finally, by using Lemma 1 and 2, the upper-bound of time derivative of  $V_4$  is calculated as

$$\dot{V}_{4}(x(t)) \leq \zeta^{T}(t) \left(h_{M}^{2} \psi^{T}(I_{N} \otimes R) \psi + \Phi_{4}\right) \zeta(t)$$
(18)

By combining (15)-(18), an upper bound of  $\dot{V}$  is obtained as:

$$\dot{V} \leq \zeta^{T}(t)(\Phi + e_{1}(I_{N} \otimes P)\psi_{2}^{T} + \psi_{2}(I_{N} \otimes P)e_{1} + h_{M}^{2}\psi^{T}(I_{N} \otimes R)\psi)\zeta(t)$$

$$(19)$$

Using Lemma 4 and Schur complement, stabilization criterion for the system (19) is equivalent to

$$\begin{bmatrix} \Xi & h_{M}\psi_{1}^{T} & e_{2}^{T} & e_{2}^{T} \\ h_{M}\psi_{1} & -(I_{N}\otimes R)^{-1} + \Psi & 0 & 0 \\ e_{2} & 0 & -\varepsilon_{1}I & 0 \\ e_{2} & 0 & 0 & -\varepsilon_{2}I \end{bmatrix} < 0$$
(20)

It should be note that the stabilization condition (20) have the non-linear term  $R^{-1}$ . So a simple method to solve it is to set  $R^{-1} = \alpha X$ , where  $\alpha > 0$  is a tuning parameter. If the LMIs (12) and (13) hold, then stability condition (11) is satisfied. This completes our proof.

## IV. NUMERICAL EXAMPLE

In this section, one numerical example will be shown to illustrate the effectiveness of the proposed Theorem 1. **Example 1.** Consider the multi-agent systems (11).

$$A = \begin{bmatrix} -1 & -2 \\ 2 & -1 \end{bmatrix}, B = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, G = \begin{bmatrix} -1 & 1 \\ 1 & -1 \end{bmatrix}$$

which satisfied with  $0 \le h(t) \le h_M$ . By applying Theorem 1, comparison with the same sampling interval  $h_M$ , when  $\underline{r} = 0.5$ ,  $\overline{r_i} = 1$  and  $\underline{r} = 1$ ,  $\overline{r_i} = 1$ . In fact, when  $\underline{r} = 1$ ,  $\overline{r_i} = 1$ , it is non-reliable systems. So we compared the reliable systems with the non-reliable systems by using Theorem 1. Their results are listed in Figs. 1-3.

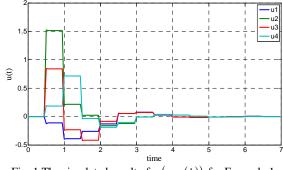


Fig. 1 The simulated result of x(t-t(h)) for Example 1

The switched interval  $h_M$  is 0.5. Then, the result is shown in Fig. 1.

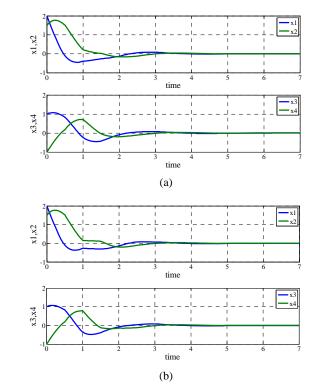


Fig. 2 The result of x(t) for example 1 when (a)  $\underline{r}_i = 0.5$ ,  $\overline{r}_i = 1$  and (b)  $r_i = 1$ ,  $\overline{r}_i = 1$ 

In order to confirm the this system result, we set that initial value of the state set up by  $x(0) = \begin{bmatrix} 0 & 0 \end{bmatrix}^T$ . The system (11) is asymptotically stable with reliable sampled-data stability.

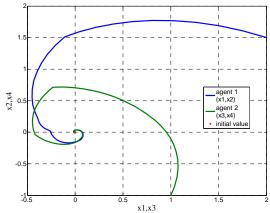


Fig. 3  $x_1(t)$ ,  $x_2(t)$  state trajectories of the systems for example 1

The figure shows that  $x_1(t)$ ,  $x_2(t)$  state are set by initial states of

$$\begin{bmatrix} x_1(0) & x_2(0) \end{bmatrix}^T = \begin{bmatrix} 2 & 1.5 \end{bmatrix}^T, \begin{bmatrix} x_3(0) & x_4(0) \end{bmatrix}^T = \begin{bmatrix} 1 & -1 \end{bmatrix}^T.$$

## V.CONCLUSION

In this paper, reliable consensus of multi-agent systems with sampled-data is proposed. To do this, constructing a suitable lemmas such as Wirtinger inequality, Reciprocally approach and Kronecker product, etc. To show the effectiveness of the proposed theorem, one numerical example was included.

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