# Average consensus and stability analysis in networked dynamic systems

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*Abstract:* - This paper provides protocols for finite-time average consensus and finite-time stability of systems with controlled nonlinear dynamics in-network under undirected fixed topology. Each node's state is a high dimensional vector as a solution of the highly nonlinear first order dynamics with and without drift terms. This paper provides protocols for finite-time average consensus and finite-time stability of systems with controlled nonlinear dynamics in-network under undirected fixed topology. Each node's state is high Under the proposed interaction rules, agreements as a common average value or an average trajectory are reached, solving finite-time average consensus and the multisystem equilibrium is controlled leading to the finite-time stability of each system origin. Sufficient conditions are achieved using the Lyapunov techniques and the graph theory. In networked dynamic systems, the theoretical results of the paper cover a large class of underactuated autonomous systems as formation flight, multi-vehicle coordination, and heterogeneous multisystem behaviors. Some examples are introduced in simulation which approves the proposed protocols.

Key-Words: - Finite-time average consensus; finite-time stability; multi-system dynamics.

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# 1 Introduction

For cooperative tasks using multi-agent groups, the presence of a large number of autonomous dynamical systems in industry requires interrelationships between distributed control parameters which are designed at a first step to manage each agent separately. Thus, in coordination of a team of autonomous agents, the communication of sensors is fundamental in many distributed control systems. For many applications the main challenges in cooperative design for a group of agents is to meet some objectives such that the rendezvous problem of multi-vehicle, control of training, flock-ing, attitude synchronization and the fusion of sensors. A coherent movement in masses is called consensus. Thus, the problem of consensus plays a central role in study of multi-agent systems. In recent years this paradigm has introduced in multiagent systems witnessed dramatic advances of various distributed strategies that achieve agreements. In [5], the authors proposed a simple but interesting discrete-time model of finite agents all moving in the plane. Each agent's motion is updated using a local rule based on its own state and the states of its neighbors. [6] provided a theoretical explanation of the consensus property of the Vicsek model by using graph theory and nonnegative matrix theory. For this model each agent's set of neighbors changes with time as system evolves. Consequently, many seemingly different problems that involve inter-connection of dynamic systems in various areas of science and engineering happen to be closely related to consensus problems for multi-agent systems. The existing connections are presented by [25] with application to linear dynamics in network in studying of multi-system behaviors.

The theoretical framework for posing and solving consensus problems for networked dynamic systems was introduced by [7] and [8]. Under dynamically changing interaction topologies, [9] extended the results of [6].

Various finite-time stabilizing control laws have been proposed using continuous state feedback and output feedback controllers [3]. Furthermore, the finite-time control design has been extended to  $n^{th}$  order systems with both parametric and dynamic uncertainties [2]. Although the finite-time design is generally more difficult than the asymptotically stabilizing control due to the lack of effective analysis tools. Also, the nonsmooth finite-time control synthesis can improve the system behaviors in some aspects like high-speed, control accuracy, and disturbance- rejection. Therefore, it is not surprising that finite-time control ideas have been applied to multi-agent systems with firstorder agent dynamics using gradient flow and Lyapunov function [10].

Finite-time consensus firstly was studied by [10], where a non-smooth consensus algorithm is proposed. In the same filed [11], and in [17] authors proposed a continuous nonlinear consensus algorithm to guaran-

tee the finite-time stability under an undirected fixed interaction graph. [16] suggest an improvement to the proposed algorithm proposed in [11]. The new algorithm proposed in [16] is able to guarantee finitetime consensus under an undirected switching interaction and a directed fixed interaction graph when each strongly connected component of the topology is detail-balanced. In [19], the authors study finite-time consensus for second order dynamics with inherent nonlinear dynamics under an undirected fixed interaction graph. In networked dynamic systems, finitetime consensus problems that have been solved so far are mostly only for simple agents like particle behaviors as first or second order dynamics. In [13] - [14], the authors treated finite-time consensus for highly nonlinear dynamic systems in network affine in control inputs. Such a system is described by a nonlinear first-order ordinary differential relations.

While an interesting topic in consensus problem is the average consensus problem such that the states of all the agents converge asymptotically or in finite time to the average of their initial states under a networked interaction protocol, one cites the results in [20] [21] [22] [23], our work consists to extend these results and propose protocols for nonlinear dynamic systems in network expected to reach an agreement that can be a predefined average value or an average trajectory. Moreover, we will make difference between consensus and stability protocols in treating the equilibrium stability of the designed multi-system dynamics.

The paper is organized as follows. Some preliminaries results, the problem statement, and the finite-time average consensus protocol are formulated in section 2. In section 3 one solves a finite- time average consensus of multi-system without drift terms. The finite-time average consensus of multi-system with drift is detailed in section 4. Finally, illustrative examples are presented in section 5.

#### 2 **Preliminaries and problem** formulation

Throughout this paper, we use R to denote the set of real number.  $\hat{R}^{\hat{n}}$  is the *n*-dimensional real vector space and ||.|| denotes the Euclidian norm.  $R^{n \times n}$ is the set of  $n \times n$  matrices. diag $\{m_1, m_2, \dots, m_n\}$  denotes a  $n \times n$  diagonal matrix.  $I_n \in \mathbb{R}^{n \times n}$  is the identity matrix. The symbol  $\otimes$  is the Kronecker product of matrices. We use sgn(.) to denote the signum function. For a scalar x, note that  $\varphi_{\alpha}(x) =$  $\operatorname{sgn}(x) \|x\|^{\alpha}$ . We use  $x^{i} = (x_{1}^{i}, x_{2}^{i}, \dots, x_{n}^{i})^{T} \in \mathbb{R}^{n}$ ,  $\mathbf{x} = (x^{1}, x^{2}, \dots, x^{N})^{T}$  to denote the vector in  $\mathbb{R}^{n \times N}$ . Let  $\phi_{\alpha}(x^{i}) = (\varphi_{\alpha}(x_{1}^{i}), \varphi_{\alpha}(x_{2}^{i}), \dots, \varphi_{\alpha}(x_{n}^{i}))^{T}$  with  $\phi_{\alpha}(\mathbf{x}) = (\phi_{\alpha}(x^{i}), \dots, \phi_{\alpha}(x^{N}))^{T}.$ Let  $\mathbf{1}_{n} = (1, \dots, 1)^{T}.$  The exponent **T** is the trans-

pose.

#### 2.1 **Graph theory**

In this subsection, we introduce some basic concepts in algebraic graph theory for multi-agent networks. Let  $\mathcal{G} = \{\mathcal{V}, \mathcal{E}\}$  be a directed graph, where

 $\mathcal{V} = \{1, 2, \dots, n\}$  is the set of nodes, node *i* represents the ith agent,  $\mathcal{E}$  is the set of edges, and an edge in  $\mathcal{G}$  is denoted by an ordered pair  $(i, j).(i, j) \in \mathcal{E}$  if and only if the ith agent can send information to the jth agent directly.  $A = [a_{ij}] \in \mathbb{R}^{n \times n}$  is called the weighted adjacency matrix of  $\mathcal{G}$  with nonnegative elements, where  $a_{ij} > 0$  if there is an edge between the ith agent and jth agent and  $a_{ij} = 0$  otherwise. Moreover, if  $A^T = A$ , then  $\mathcal{G}$  is also called an undirected graph. In this paper, we will refer to graphs whose weights take values in the set  $\{0, 1\}$  as binary and those graphs whose adjacency matrices are symmetric as symmetric.

Let  $D = \text{diag}\{d_1, \dots, d_n\} \in \mathbb{R}^{n \times n}$  be a diagonal matrix, where  $d_i = \sum_{j=1}^n a_{ij}$  for  $i = 0, 1, \dots, n$ . Hence,

we define the Laplacian of the weighted graph

$$L = D - A \in R^{n \times n}$$

The undirected graph is called connected if there is a path between any two vertices of the graph.

#### Some useful lemmas 2.2

Our main results are guided by the following Lemmas. The reader may find more details in the associated references.

Lemma 1 : Bhat and Bernstein(200). Consider the system  $\dot{x} = f(x), f(0) = 0, x \in \mathbb{R}^n$ , there exist a positive definite continuous function

 $V(x) : U \subset \mathbb{R}^n \to \mathbb{R}$ , real numbers c > 0 and  $\alpha \in [0,1[$ , and an open neighborhood  $U_0 \subset U$  of the origin such that

 $\dot{V} + c(V(x))^{\alpha} \leq 0, x \in U_0 \setminus \{0\}$ . Then V(x) converges to zero in finite time. In addition, the finite settling time  $T_{\star}$  satisfies  $T_{\star} \leq \frac{V((x(0))^{1-\alpha}}{c(1-\alpha)}$ .

**Lemma 2** : Hong & al (2002). Consider the follow-ing system,  $x = [x_1, \ldots, x_n]^T \in \mathbb{R}^n$ 

$$\dot{x} = g(x) + \tilde{g}(x) \tag{1}$$

where g(0) = 0 and g(x) is a continuous homogeneous vector field of degree d < 0 with respect to dilation  $[\sigma_1, \ldots, \sigma_n]$ , and  $\tilde{g}(x) = [\tilde{g}_1(x), \ldots, \tilde{g}_n(x)]^T \in$  $R^n$  satisfies  $\tilde{g}(0) = 0$ . Assume that x = 0 is an asymptotically stable equilibrium of the system  $\dot{x} =$ g(x). Then x = 0 is a globally finite time stable equi*librium of system* (1) *if* 

**Lemma 3** : Olfati-Saber & al (2004). For a connected undirected graph  $\mathcal{G}$ , the Laplacian matrix L of  $\mathcal{G}$  has he following properties,

$$x^T L x = \frac{1}{2} \sum_{i,j=1}^{n} a_{ij} (x_i - x_j)^2$$
, which implies that L is

positive semi-definite. 0 is a simple eigenvalue of Land  $\mathbf{1}$  is the associated eigenvector. Assume that the eigenvalues of L are denoted by  $0, \lambda_2, \ldots, \lambda_n$  satisfying  $0 \le \lambda_2 \le \cdots \le \lambda_n$ . Then the smallest eigenvalue satisfies  $\lambda_2 > 0$ . Furthermore, if  $\mathbf{I}^T x = 0$ , then  $x^T L x \ge \lambda_2 x^T x$ .

Lemma 4 : Hardy & al (1952). Let  $x_1, x_2, ..., x_n \ge 0$  and o . Then $<math>\left(\sum_{i=1}^n x_i\right)^p \le \sum_{i=1}^n x_i^p \le n^{1-p} \left(\sum_{i=1}^n x_i\right)^p$ .

### 2.3 **Problem statements**

We solve the finite-time average consensus and stability of two type of models in networked dynamic systems affine in control inputs. The first type is given by equation (2) which describes a controlled dynamic system without drift term. The second type is represented by relation (3) which is clearly a controlled dynamic system with drift term  $f^i(x_i)$ . Let consider a group of N high-dimensional agents where each agent's behavior is described by a controlled nonlinear model without drift  $\Sigma_1$  represented by the controlled dynamic (2) and system  $\Sigma_2$  with drift as shown by the controlled dynamic (3),  $\forall i \in \mathcal{I} = \{1, \ldots, N\}$ 

$$\Sigma_1 : \dot{x^i} = B(x^i)u^i. \tag{2}$$

and

$$\Sigma_2 : \dot{x^i} = f^i(x^i) + B(x^i)u^i.$$
 (3)

where  $x^i \in R^n, x^i = [x_1^i, x_2^i, \dots, x_n^i]^T, B(x^i) \in R^{n \times m}$ , the continuous maps  $f^i : R^n \to R^n, u^i \in R^m$  is the control input and for  $1 \le k \le n$  and  $1 \le m, B(x^i) = [b_{kl}]$ .

**Definition 1** (stabilization) Given an interconnection control  $u^i(x^i, x^j)$ , the origin the zero solution  $x^i(t) = 0$  to (2)-(3) is finite-time stable if the following statements hold:

- *1.* The zero solution of closed loop system to (2)-(3) is stable.
- 2. There exist a settling-time  $T_{\star}$  such that  $\lim_{t \to T_{\star}} ||x^{i(t)}|| = 0$

**Definition 2** Given a control-input  $u^i$  as protocol, we say that systems in network meet a finite- time average consensus if for any system's state initial conditions, there exists some finite time  $T_*$  such that:

$$\lim_{t \to T_{\star}} ||x^{i(t)} - \chi(t)|| = 0$$
(4)

for any  $i \in \mathcal{I}$ , and where  $\chi(t) = \frac{1}{N} \sum_{j=1}^{N} x^{j}(t)$  is the

average trajectory.

 $\chi(t)$  can be interpreted as the instantaneous consent providing that serves the group objectives.  $\chi$  is timevarying, it can be also considered as the average trajectory of the group, and it is not necessary the average from the multi-system initial conditions. We show that the dynamic of  $\chi$  depends strongly on the adopted topology of the group.

Subsequently, for the multi- $\Sigma_1$  and multi- $\Sigma_2$  systems one might analyze the following protocols are given by (5)and (6). For  $i \in \mathcal{I}$ , the consensus protocol candidate is given by,

$$u^{i} = -C(x^{i}) \sum_{j=1}^{N} a_{ij} \phi_{\alpha}(x^{i} - x^{j})$$
 (5)

while the stabilizing input candidate is as

$$u^{i} = -C(x^{i}) \sum_{j=1}^{N} a_{ij}(\phi_{\alpha}(x^{i}) - \phi_{\alpha}(x^{j}))$$
 (6)

where the  $a_{ij}$  elements are of the  $\mathcal{G}$  adjacency matrix,  $\alpha \in ]0, 1[$ , and  $\phi_{\alpha}(.)$  is defined in section 2. The control matrix  $C(x^i) \in \mathbb{R}^{m \times n}$  depends on the agent's model, and it will be defined in the following.

As we can see in protocols (5) and (6), the finitetime average consensus is closely related to finitetime stability. The main difference between the two problems is that finite-time average consensus is to make the multi-system converge to an agreement value or trajectory as given by  $\chi(t)$  in (4), while the stability of each agent consists to reach an equilibrium. The following assumption gives a conceptual form of  $C(x^i)$  with respect to the studied dynamics. **Assumption1:**  $C(x^i)$  is such that the matrix product  $B(x^i)C(x^i)$  is positive semi-definite matrix.

Throughout the paper, one denotes by

 $B(x^i) = B(x^i)C(x^i).$ Assumption 2: For a given control

Assumption 2: For a given control matrix  $C(x^i)$ , for all  $x, y \in \mathbb{R}^n$ , we assume that

$$(\phi_{\alpha}(x) - \phi_{\alpha}(y))^{T} \tilde{B}(x^{i}) \phi_{\alpha}(x - y) \geq (\phi_{\alpha}(x) - \phi_{\alpha}(y))^{T} \tilde{B}(x^{i}) (\phi_{\alpha}(x) - \phi_{\alpha}(y))$$
*Assumption3: Consider that*

$$g(x^i) = -\sum_{j=1}^N a_{ij}\tilde{B}(x^i)\phi_\alpha(\xi^i - \xi^j)$$
(7)

is a locally homogeneous vector field of degree d with respect to dilation  $[\sigma_1, \ldots, \sigma_n]$ .

#### 3 Finite-time average consensus

The objective of this section is to solve finite-time average consensus problems of multi-system based on  $\Sigma_1$  and  $\Sigma_2$  descriptions. The average value is considered as an agreement function of time but is not necessary function of multi-agent initial conditions. Further, what motivates the analysis is that models given by (3) and (2) cover many autonomous system behaviors affine in the control vector. One may cite, automated highway systems, multi-drone, multi-system of satellites or robots, etc. When we refer to the protocol(5), the interaction topology uses undirected flow information between nodes where each node's vector of states is as a solution of (3) or (2). The following two subsections treat the multi- $\Sigma_1$  and multi- $\Sigma_2$ finite-time average consensus.

#### The multi- $\Sigma_1$ finite-time average 3.1 consensus

For finite-time average consensus of multi- $\Sigma_1$  one considers, as interaction topology an undirected fixed graph, an average vector obtained from each  $\Sigma_1$  vector of states, and the protocol candidate (5). As the matrix B structure is taken identical for each  $\Sigma_1$  then one might think to networked homogeneous systems. Recall that for a group where each agent is of the form  $\dot{x}^i = u^i$ , if the interconnection topology is based on an undirected flow, then the average consensus is solved with respect to the average of the agents initial states.

**Proposition 1** Let *G* be an undirected and connected graph, under the protocol (5) and Assumptions 1-2-3 the multi- $\Sigma_1$  achieves a finite-time average consensus in the sense of (4).

**Proof** We introduce  $\xi^i(t) = x^i(t) - \chi(t)$  and  $\xi(t) = [\xi^1, ..., \xi^N]^T$ . Due to the fact that  $a_{ij} = a_{ji}$  for all  $1 \le i, j \le N$  (undirected graph) and  $\varphi_{\alpha}$  is an odd function, we have,

$$\begin{split} \dot{\chi}(t) &= \frac{1}{N} \sum_{i=1}^{N} \dot{x}^{i}(t) \\ &= -\frac{1}{N} \sum_{i,j=1}^{N} a_{ij} \tilde{B}(x^{i}) \phi_{\alpha}(x^{i} - x^{j}) \\ &= -\frac{1}{2N} \sum_{i,j=1}^{N} a_{ij} \left( \tilde{B}(x^{i}) - \tilde{B}(x^{j}) \right) \phi_{\alpha}(x^{i} - x^{j}) \end{split}$$

Introducing the protocol (5), we obtain

(8)

$$\begin{split} \xi^{i}(t) &= x^{i}(t) - \chi(t) \\ &= -\sum_{j=1}^{N} a_{ij} \tilde{B}(x^{i}) \phi_{\alpha}(x^{i} - x^{j}) + \\ &\frac{1}{2N} \sum_{i,j=1}^{N} a_{ij} (\tilde{B}(x^{i}) - \tilde{B}(x^{j})) \phi_{\alpha}(x^{i} - x^{j}) \\ &= -\sum_{j=1}^{N} a_{ij} \tilde{B}(x^{i}) \phi_{\alpha}(\xi^{i} - \xi^{j}) + \\ &\frac{1}{2N} \sum_{i,j=1}^{N} a_{ij} (\tilde{B}(x^{i}) - \tilde{B}(x^{j})) \phi_{\alpha}(\xi^{i} - \xi^{j}) \\ Let \ \xi(t) &= (\xi^{1}, ..., \xi^{N}), \ we \ can \ write \ the \ last \ equation \ in \ the \ form: \\ & \dot{\xi}^{i}(t) = g(\xi^{i}) + \tilde{g}(\xi) \end{split}$$
(8)

where

and

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$$g(\xi^i) = -\sum_{j=1}^N a_{ij}\tilde{B}(x^i)\phi_\alpha(\xi^i - \xi^j)$$

$$\tilde{g}(\xi) = \frac{1}{2N} \sum_{i,j=1}^{N} a_{ij} \left( \tilde{B}(x^i) - \tilde{B}(x^j) \right) \phi_{\alpha}(\xi^i - \xi^j)$$

By now, it remains to prove that the equilibrium of (8) is finitetime stable, and this is achieved in the subsequent two steps. Step 1. First, the goal is to prove the finite time stability of system

$$\dot{\xi}^i(t) = g(\xi^i) \tag{9}$$

Taking the Lyapunov function:

$$V(\xi(t)) = \frac{1}{\alpha + 1} \sum_{i=1}^{N} (\xi^{i})^{T} \phi_{\alpha}(\xi^{i})$$
(10)

The derivative of V along the solutions of system (9), yields

$$\begin{split} \dot{V}(\xi(t)) &= \sum_{i=1}^{N} \left( \phi_{\alpha}(\xi^{i}) \right)^{T} \dot{\xi}^{i} \\ &= -\sum_{i,j=1}^{N} a_{ij} \left( \phi_{\alpha}(\xi^{i}) \right)^{T} \tilde{B}(x^{i}) \phi_{\alpha}(\xi^{i} - \xi^{j}) \\ &= -\frac{1}{2} \sum_{i,j=1}^{N} a_{ij} \left( \phi_{\alpha}(\xi^{i}) - \phi_{\alpha}(\xi^{j}) \right)^{T} \tilde{B}(x^{i}) \phi_{\alpha}(\xi^{i} - \xi^{j}) \end{split}$$
From Assumption 2, the following inequality holds,

$$\begin{split} \dot{V} &\leq -\frac{1}{2} \sum_{i,j=1}^{N} a_{ij} (\phi_{\alpha}(\xi^{i}) - \phi_{\alpha}(\xi^{j}))^{T} \tilde{B}(x^{i}) \left(\phi_{\alpha}(\xi^{i}) - \phi_{\alpha}(\xi^{j})\right) \\ &= -\frac{1}{2} \phi_{\alpha}^{T}(\xi) \left(L \otimes \tilde{B}(x^{i})\right) \phi_{\alpha}(\xi) \\ &= -\frac{1}{4} \phi_{\alpha}^{T}(\xi) \Theta \phi_{\alpha}(\xi) \\ &\text{where } \Theta = \frac{1}{2} \left(L \otimes \tilde{B}(x^{i}) + L \otimes \tilde{B}^{T}(x^{i})\right). \end{split}$$

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 $D(x^{i}) = diag\{0_{n}, \gamma_{2}(x^{i}), ..., \gamma_{N}(x^{i})\}$ such that  $0_n = diag\{0, ..., 0\} \in \mathbb{R}^{n \times n}$  and  $\forall j = 2, ..., N$  $\gamma_j(x^i) = \lambda_j(L)\varrho_n(x^i)$  where

 $\varrho_n(x^i) = diag\{0, \mu_2(x^i), ..., \mu_n(x^i)\} \in \mathbb{R}^{n \times n}$  and where  $\mu_2(x^i), ..., \mu_n(x^i)$  are the eigenvalues of the matrix  $\tilde{B}(x^i)$ , given in increasing order:  $\lambda_j(L)$  is the  $j^{\text{ème}}$  eigenvalue of L. Let  $\lambda_2(L), ..., \lambda_N(L)$  in increasing order. Since G is connected (by Lemma 3)  $\lambda_2(L) > 0$ . Therefore  $\forall x^i$ , we have  $\lambda_2\mu_2(x^i) > 0$ . Further, since  $\Theta \in \mathbb{R}^{Nn \times Nn}$  is symmetric matrix, then there exist an orthogonal matrix  $P \in \mathbb{R}^{Nn \times Nn}$  such that  $\Theta = P^T D(x^i)P$ . Let  $z_\alpha = P\phi_\alpha(\xi)$ , thus

$$\dot{V} \leq -\frac{1}{4} z_{\alpha}^{T} D z_{\alpha} \\ \leq -\frac{1}{4} \lambda_{2} \mu_{1}(x^{i}) \| z_{\alpha} \|^{2} \\ = -\frac{1}{4} \lambda_{2} \mu_{1}(\xi^{i}) \| \phi_{\alpha}(\xi) \|^{2}$$

where  $\lambda_2 \mu_1(x^i) = \min_{z_\alpha \perp I_{N_n}} \frac{z_\alpha^T D z_\alpha}{z_\alpha^T z_\alpha}$ . Let  $k = \min_{x^i \in \mathbb{R}^N} \lambda_2 \mu_1(x^i) > 0$  and  $\xi = I_N \otimes \xi^i = (\tilde{\xi}_1, ..., \tilde{\xi}_{N_n})^T$ , consequently,

$$\dot{V} \leq -\frac{k}{4} \sum_{i=1}^{Nn} |\varphi_{\alpha}(\tilde{\xi}_{i})|^{2} \\
\leq -\frac{k}{4} \sum_{i=1}^{Nn} |\tilde{\xi}_{i}|^{2\alpha} \\
\leq -\frac{k}{4} \left( \sum_{i=1}^{Nn} |\tilde{\xi}_{i}|^{\alpha+1} \right)^{\frac{2\alpha}{\alpha+1}}$$
(11)

which permits to write

$$\dot{V} \le -\frac{k}{4}(\alpha+1)^{\frac{2\alpha}{\alpha+1}}V^{\frac{2\alpha}{\alpha+1}} \tag{12}$$

where  $0 < \frac{2\alpha}{\alpha+1} < 1$  and  $\frac{k}{4}(\alpha+1)^{\frac{2\alpha}{\alpha+1}} > 0$ , by Lemma 1, the above differential equation (9) shows that V reaches zero in finite time..

**Step 2**. From Assumption 3, the vector field  $g(\xi^i)$  is homogeneous of degree d which is negative due to the fact that  $\xi = 0$  is a finite time stable equilibrium. Moreover, it is straightforward to prove that  $\tilde{g}(\xi)$ , pour k = 1, ..., n, il est simple de vérifier que

$$\lim_{\varepsilon \to 0} \frac{\tilde{g}_k(\varepsilon^{\sigma_1} \xi_1^i, ..., \varepsilon^{\sigma_n} \xi_n^i)}{\varepsilon^{d+\sigma_i}} = 0$$

Then by Lemma ??, the system (8) is finite time stable. Thus, as a result the multi- $\Sigma_1$  dynamic system with the protocol (5) solve a finite-time average consensus. This ends the proof.

# **3.2** The multi- $\Sigma_2$ finite-time average consensus

The multi- $\Sigma_2$  behavior is based on (3) while the consensus protocol candidate is given by ((5). Recall that the  $\Sigma_2$  dynamic as given by (3) is currently present in controlled autonomous systems. However, the drift term can be linear with respect to the system's state vector or taken in its nonlinear form. These two issues will be analyzed in the following with the adequate sufficient conditions for multi- $\Sigma_2$  finite-time average consensus. To do, let us first note that fi in (3) can be different for each dynamic leading to heterogeneous multi-system. At first, the subsequent analysis is build on this form of  $f^i(x^i) \triangleq \tilde{A}x^i$  with  $\tilde{A}$  is a constant matrix. A controlled dynamic system with linear drift term is given by,

$$x^{i} = \tilde{A}x^{i} + B(x^{i})u^{i} \tag{13}$$

where  $\tilde{A} \in \mathbb{R}^{n \times n}$  with  $\tilde{A} = [\tilde{a}_{p,q}]_{1 \le p,q \le n}$ .

**Proposition 2** Let  $\mathcal{G}$  be an undirected and connected graph, under the protocol (5) the multi- $\Sigma_2$ , built from (13), converges toward an average trajectory and leads to a finite-time average consensus in the sense of (4).

**Proof** One introduces  $\xi^i(t) = x^i(t) - \chi(t)$ . The goal is to rewrite equation (13) in closed loop depending on  $\xi^i$  and to prove that  $\xi$  converges to zero in finite time. Since  $a_{ij} = a_{ji}$  and  $\phi_{\alpha}(.)$  is an odd function, then we have

$$\begin{split} \dot{\chi}(t) &= \frac{1}{N} \sum_{i=1}^{N} (\tilde{A}x^{i} + B(x^{i})u^{i}) \\ &= \frac{1}{N} \sum_{i=1}^{N} \tilde{A}x^{i} + \frac{1}{N} \sum_{i=1}^{N} B(x^{i})u^{i} \\ &= \frac{1}{N} \sum_{i=1}^{N} \tilde{A}x^{i} - \frac{1}{2N} \sum_{i,j=1}^{N} a_{ij}(\tilde{B}(x^{i}) - \tilde{B}(x^{j}))\phi_{\alpha}(x^{i} - x^{j}) \\ &\tilde{\xi}^{i} = \tilde{A}\xi^{i} - \sum_{j=1}^{N} a_{ij}\tilde{B}(x^{i})\phi_{\alpha}(\xi^{i} - \xi^{j}) + \\ &\frac{1}{2N} \sum_{i,j=1}^{N} a_{ij}(\tilde{B}(x^{i}) - \tilde{B}(x^{j}))\phi_{\alpha}(\xi^{i} - \xi^{j}) \end{split}$$

keeping the same steps of the previous proof, we introduce

$$\dot{\xi}^i = h(\xi^i) + \tilde{h}(\xi)$$

where

$$h(\xi^i) = \tilde{A}\xi^i - \sum_{j=1}^N a_{ij}\tilde{B}(x^i)\phi_\alpha(\xi^i - \xi^j)$$

$$\tilde{h}(\xi) = \frac{1}{2N} \sum_{i,j=1}^{N} a_{ij} (\tilde{B}(x^{i}) - \tilde{B}(x^{j})) \phi_{\alpha}(\xi^{i} - \xi^{j})$$

where  $h(\xi)\tilde{g}(\xi)$  then it remains to prove the finite-time stability of the system.

$$\dot{\xi}^i = h(\xi^i) \tag{14}$$

Using the Lyapunov function (10), the time derivative of  $V(\xi)$  along the networked system trajectories (14) is given by

$$\begin{split} \dot{V}(\xi(t)) &= \sum_{i=1}^{N} \phi_{\alpha}^{T}(\xi^{i}) \dot{\xi}^{i} \\ &= \sum_{i=1}^{N} \phi_{\alpha}^{T}(\xi^{i}) \tilde{A}\xi^{i} - \sum_{i,j=1}^{N} a_{ij} \phi_{\alpha}^{T}(\xi^{i}) \tilde{B}(x^{i}) \phi_{\alpha}(\xi^{i} - \xi^{j}) \\ &\leq \|\tilde{A}\|_{\infty} \sum_{i=1}^{N} \phi_{\alpha}^{T}(\xi^{i}) \xi^{i} - \frac{k}{4} (\alpha + 1)^{\frac{2\alpha}{\alpha+1}} V^{\frac{2\alpha}{\alpha+1}} \\ &\leq \|\tilde{A}\|_{\infty} V(\xi(t)) - \frac{k}{4} (\alpha + 1)^{\frac{2\alpha}{\alpha+1}} V^{\frac{2\alpha}{\alpha+1}} \\ &\leq -V(\xi(t))^{\frac{2\alpha}{\alpha+1}} [\frac{k}{4} (\alpha + 1)^{\frac{2\alpha}{\alpha+1}} - \|\tilde{A}\|_{\infty} (V(\xi(t)))^{\frac{1-\alpha}{\alpha+1}}] \\ &\qquad \text{where } \|\tilde{A}\|_{\infty} = \max_{1 \leq p \leq n} \sum_{q=1}^{n} |\tilde{a}_{pq}| > 0. \quad \text{Since} \end{split}$$

 $\frac{1-\alpha}{\alpha+1} > 0$  and V is continuous function which takes  $(V(\xi) = 0)$  there exists an open neighborhood  $\Omega$  of the origin such that the last inequality

$$\dot{V}(\xi(t)) \le -\frac{k}{8}(\alpha+1)^{\frac{2\alpha}{\alpha+1}}V(\xi(t))^{\frac{2\alpha}{\alpha+1}}$$
 (15)

By Lemma 1, V reaches zero in finite time. Therfore  $\xi^i = 0$  is a finite-time stable equilibrium of system (14) We may follow step 2 of the previous analysis to end the proof.

In the following, we consider that the drift term in (3) is nonlinear which also commonly present in controlled dynamic systems. Moreover, if the networked dynamic systems is homogenous then

the  $f^i$  structure is identical, otherwise the multisystem is considered as heterogenous. Our main result in multi- $\Sigma_2$  is built on the assumption that  $f^i(x^i)$ is a convex function.

**Proposition 3** Let  $\mathcal{G}$  be a fixed undirected graph and  $f^i(x^i)$  is convex. Under the protocol (5) a homogenous/heterogenous multi- $\Sigma_2$  based on (3) converges toward an average trajectory and leads to a finite-time average consensus in the sense of (4).

**Proof** Let  $\xi^i(t) = x^i(t) - \chi(t)$ . As  $f^i$  is assumed to be convex, we have

$$f^{i}(x^{i}) - \frac{1}{N} \sum_{i=1}^{N} f^{i}(x^{i}) \le f^{i}(x^{i}) - f^{i}\left(\frac{1}{N} \sum_{i=1}^{N} x^{i}\right)$$

Moreover  $f^i$  is locally Lipschitz function in an open set  $\Omega \subset \mathbb{R}^n$  containing  $\xi$ . Therefore

$$f^{i}(x^{i}) - \frac{1}{N} \sum_{i=1}^{N} f^{i}(x^{i}) \leq \|f^{i}(x^{i}) - f^{i}(\chi)\|$$
$$\leq c \|\xi^{i}\|$$

such that c > 0 is the Lipschitz's constant. Now, for convenience the Lyapunov function is given by (10), we prove:

$$\begin{split} \dot{V}(\xi(t)) &= \sum_{i=1}^{N} (\phi_{\alpha}(\xi^{i}))^{T} \dot{\xi}^{i} \\ &\leq c \sum_{i=1}^{N} \phi_{\alpha}^{T}(\xi^{i}) \xi^{i} - \frac{k}{4} (\alpha + 1)^{\frac{2\alpha}{\alpha+1}} V^{\frac{2\alpha}{\alpha+1}} \\ &\leq -V(\xi(t))^{\frac{2\alpha}{\alpha+1}} \left[ \frac{k}{4} (\alpha + 1)^{\frac{2\alpha}{\alpha+1}} - c(V(\xi(t)))^{\frac{1-\alpha}{\alpha+1}} \right] \end{split}$$

*Or*  $\frac{1-\alpha}{\alpha+1} > 0$  *and V is a continuous function which takes* 0 *of the origin* V(0) = 0 *there exists an open neighborhood*  $\Omega$  *such that*  $\xi(t) \in \Omega$ 

$$\dot{V}(\xi(t)) \le -\frac{k}{8}(\alpha+1)^{\frac{2\alpha}{\alpha+1}}V(\xi(t))^{\frac{2\alpha}{\alpha+1}}$$
 (16)

At this stage, one concludes that the multi- $\Sigma_2$  established from with the protocol (3) with the protocol (5) lead to a finite-time average consensus. Note that if the convexity property of  $f^i$  is not satisfied, the alternative is to linearize each  $\Sigma_2$  system and use the same procedure obtained for a multi-system built from (13).

# 4 The multi-system finite-time stabilization

The finite-time stabilization problem in networked dynamic systems consists to stabilize individually each system's equilibrium state under some connection rules. Then we consider dynamic systems in network with continuous nonlinear decentralized feedback that integrates the graph theory. The following theoretical framework tackles first to the multi- $\Sigma_1$  stabilization problem, the results will be extended after that to the analysis of the multi- $\Sigma_2$  stabilization problem.

#### **4.1** The multi- $\Sigma_1$ finite-time stabilization

The multi- $\Sigma_1$  describes the behavior of drift less systems like kinematic of unicycles and attitude of satellites. Further, one considers here that each system is nonlinear and not necessary fully actuated (dimension of the input vector is fewer than the system degree of freedom).

**Proposition 4** For a given fixed underacted graph G, the protocol (6) applied to multi- $\Sigma_1$  solves the stabilizing problem in finite time.

**Proof** Let  $\mathbf{x} = (x^1, ..., x^N)^T \in \mathbb{R}^{Nn}$  and  $\mathbf{u} = (u^1, ..., u^N)^T \in \mathbb{R}^{Nm}$  where  $x^i \in \mathbb{R}^n$  and  $u^i \in \mathbb{R}^m$ . The networked systems (2) under the stabilizing protocol (17)

$$\boldsymbol{u} = -(L \otimes I_n)(I_N \otimes C(x^i))\phi_\alpha(\boldsymbol{x})$$
(17)

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Using the Kronecker product properties

$$\dot{\mathbf{x}} = (I_N \otimes B(x^i))\mathbf{u} 
= -(I_N \otimes B(x^i))(L \otimes I_n)(I_N \otimes C(x^i))\phi_\alpha(\mathbf{x}) 
= -(L \otimes \tilde{B}(x^i))\phi_\alpha(\mathbf{x})$$
(18)

It is obvious from (18) that the equilibrium is zero. The goal is to prove that  $\mathbf{x}$  reaches this equilibrium in finite time. Taking the Lyapunov function  $V : \mathbb{R}^{Nn} \to \mathbb{R}_+$  where  $\forall \mathbf{x} \in \mathbb{R}^{Nn}$ 

$$V(\mathbf{x}) = \frac{1}{1+\alpha} \mathbf{x}^T \phi_\alpha(\mathbf{x})$$
(19)

which is positive definite with respect to x, Now, the time derivative along the trajectories of (18) lead to

$$\begin{split} \dot{V}(\mathbf{x}) &= \phi_{\alpha}^{T}(\mathbf{x}) \frac{d\mathbf{x}}{dt} \\ &= -\phi_{\alpha}^{T}(\mathbf{x}) (L \otimes \tilde{B}) \phi_{\alpha}(\mathbf{x}) \end{split}$$

Let

$$D(x^i) = \begin{bmatrix} 0_n & & & \\ & \gamma_2(x^i) & & \\ & & \ddots & \\ & & & & \gamma_N(x^i) \end{bmatrix}$$

where  $0_n = diag\{0, ..., 0\} \in \mathbb{R}^{n \times n}$  and  $\forall j = 2, ..., N, \ \gamma_j(x^i) = \lambda_j(L)\varrho_n(x^i)$  where  $\varrho_n(x^i) = diag\{0, \mu_2(x^i), ..., \mu_n(x^i)\} \in \mathbb{R}^{n \times n}$ . Denotes that  $\mu_2(x^i), ..., \mu_n(x^i)$  are the eigenvalues of the matrix  $\tilde{B}(x^i)$ , given in increasing order.  $\lambda_j(L)$ is the  $j^{\text{eme}}$  eigenvalues of L. Let  $\lambda_2(L), ..., \lambda_N(L)$  in increasing order. By Lemma 3,  $\lambda_2(L) > 0$ . We have  $\forall x^i, \lambda_2\mu_2(x^i) > 0$ .

Further, since  $L \otimes \tilde{B} \in \mathbb{R}^{Nn \times Nn}$  is symmetric matrix, then there exist an orthogonal matrix  $P \in \mathbb{R}^{Nn \times Nn}$ such that  $L \otimes \tilde{B} = P^T D(x^i) P$ . Let  $\mathbf{z}_{\alpha} = P \phi_{\alpha}(\mathbf{x})$ . Then

$$\begin{split} \dot{V} &= -\boldsymbol{z}_{\alpha}^{T} D \boldsymbol{z}_{\alpha} \\ &\leq -\lambda_{2} \mu_{1}(x^{i}) \|\boldsymbol{z}_{\alpha}\|^{2} \\ &\leq -\lambda_{2} \mu_{1}(x^{i}) \|\phi_{\alpha}(\boldsymbol{x})\|^{2} \end{split} \tag{20}$$

where

$$\lambda_2 \mu_1(x^i) = \min_{\mathbf{z}_{\alpha} \perp \mathbf{I}_{Nn}} \frac{\mathbf{z}_{\alpha}^T D \mathbf{z}_{\alpha}}{\mathbf{z}_{\alpha}^T \mathbf{z}_{\alpha}}$$

Let  $k = \min_{x^i \in \mathbb{R}^N} \lambda_2 \mu_1(x^i) > 0$  and

$$\begin{aligned} \mathbf{x} &= \mathbf{I}_N \otimes x^i = (\tilde{x}_1, ..., \tilde{x}_{Nn})^T, \text{ we obtain} \\ \dot{V} &\leq -k \sum_{i=1}^{Nn} |\varphi_\alpha(\tilde{x}_i)|^2 \\ &\leq -k \sum_{i=1}^{Nn} |\tilde{x}_i|^{2\alpha} \\ &\leq -k \left(\sum_{i=1}^{Nn} |\tilde{x}_i|^{\alpha+1}\right)^{\frac{2\alpha}{\alpha+1}} by \text{ Lemma 4, } \end{aligned}$$

which leads to

$$\dot{V} \le -k(\alpha+1)^{\frac{2\alpha}{\alpha+1}}V^{\frac{2\alpha}{\alpha+1}} \tag{22}$$

Or  $0 < \frac{2\alpha}{\alpha+1} < 1$  et  $k(\alpha+1)^{\frac{2\alpha}{\alpha+1}} > 0$ , by Lemma 1 the above differential equation shows that V reaches zero in finite time

$$T_*(\mathbf{x}(0)) = \frac{(\alpha+1)V(\mathbf{x}(0))^{\frac{1-\alpha}{\alpha+1}}}{(1-\alpha)k(\alpha+1)^{\frac{2\alpha}{\alpha+1}}}$$

Therefore, based on (2), the multi- $\Sigma_1$  under the protocol (6) reaches zero in finite-time.

### **4.2** The multi- $\Sigma_2$ finite-time stabilization

Recall that the multi- $\Sigma_2$  system is based on the following dynamic with nonlinear drift terms

$$\Sigma_2 : \dot{x^i} = f^i(x^i) + B(x^i)u^i.$$
(23)

where the  $f^i$  structure can be taken different for each system. In this case, we are in presence of heterogeneous multi-system. We assume at first that

$$\phi_{\alpha}^{T}(x^{i})f^{i}(x^{i}) \leq 0.$$
(24)

and we propose the following,

**Proposition 5** Suppose that the inequality (24) is satisfied. For a given fixed underacted and connected graph  $\mathcal{G}$ , the protocol (6) associated to multi- $\Sigma_2$ solves the stabilizing problem in finite time.

**Proof** Let  $\mathbf{x} \in \mathbb{R}^{Nn}$  and  $f(\mathbf{x}) = (f^1(x^1), ..., f^N(x^N))^T$ . Consider the stabilizing protocol (17), the multi- $\Sigma_2$  dynamic becomes:

$$\dot{\boldsymbol{x}} = f(\boldsymbol{x}) - (L \otimes \tilde{B})\phi_{\alpha}(\boldsymbol{x})$$
(25)

Using the Lyapunov function (19), its time derivative is as

$$\dot{V}(\boldsymbol{x}) = \phi_{\alpha}^{T}(\boldsymbol{x})f(\boldsymbol{x}) - \phi_{\alpha}^{T}(\boldsymbol{x})(L \otimes \tilde{B})\phi_{\alpha}(\boldsymbol{x}) \quad (26)$$

From hypothesis (24), the first term in (26) is negative. The remaining terms in (26) must verify the inequality given by (22). So, we conclude that the origin (25) is finite-time stable. **Remark 1** In practice condition (24) on the drift term isn't often verified. For this propose this condition can be relaxed by the following proposition.

**Proposition 6** If  $f^i$  is locally Lipshitz function and  $f^i(\boldsymbol{\theta}_n) = \boldsymbol{\theta}_n$ , given an underacted and connected graph  $\mathcal{G}$ , the multi- $\Sigma_2$  origin from (23) and (6) is locally finite-time stable.

**Proof** *Recall that the time derivative of the Lyapunov candidate function (19)* 

$$\dot{V}(\mathbf{x}) = \phi_{\alpha}^{T}(\mathbf{x})f(x) - \phi_{\alpha}^{T}(\mathbf{x})(L \otimes \tilde{B})\phi_{\alpha}(\mathbf{x}) \\
\leq c \|\phi_{\alpha}^{T}(\mathbf{x})\mathbf{x}\| - \phi_{\alpha}^{T}(\mathbf{x})(L \otimes \tilde{B})\phi_{\alpha}(\mathbf{x})$$
(27)

where c > 0 is the Lipshitz's constant.

Let  $\mathbf{x} = \mathbf{1}_N \otimes x^i = (\tilde{x_1}, \dots, \tilde{x_{Nn}})^T$ , consequently from (21), the inequality (27) permits to write

$$\dot{V}(\mathbf{x}) \leq c \left(\sum_{i=1}^{Nn} |\tilde{x}_i|^{\alpha+1}\right) - k \left(\sum_{i=1}^{Nn} |\tilde{x}_i|^{\alpha+1}\right)^{\frac{2\alpha}{\alpha+1}} \leq -V^{\frac{2\alpha}{\alpha+1}} [k(1+\alpha)^{\frac{2\alpha}{\alpha+1}} - cV^{\frac{1-\alpha}{\alpha+1}}]$$
(28)

where  $k = \min_{x^i \in R^N} \lambda_2 \mu_1(x^i)$  defined in the proof of Proposition 4. Since  $\frac{1-\alpha}{\alpha+1} > 0$  and V is continuous function which takes 0 at the origin, there exists an open neighborhood  $\Omega \subset R^{Nn}$  of the origin that permits to write

$$\dot{V}(\boldsymbol{x}) \leq -k \frac{(\alpha+1)^{\frac{2\alpha}{\alpha+1}}}{2} [V(\boldsymbol{x})]^{\frac{2\alpha}{\alpha+1}}$$
(29)

by Lemma 1, V reaches zero at an estimated finite time

$$T_*(\mathbf{x}(0)) = \frac{(\alpha+1)V(\mathbf{x}(0))^{\frac{1-\alpha}{\alpha+1}}}{2(1-\alpha)k(\alpha+1)^{\frac{2\alpha}{\alpha+1}}}$$

Therefore, based on (23) and (6), the multi- $\Sigma_2$  origin is finite-time stable.

From the proposed stabilizing protocol, we may conclude that the stability of each agent was asserted from the networked behavior of the group. Further, the drift term is not present in the protocol, however along the proofs, this term is tackled by the control and sufficient conditions on this term were introduced to guarantee the multi-system stability. Note that in individual dynamic system stability problem, the drift term must be compensated by the control-input. Here, the stability of each agent is obtained from the stable behavior of the group. This analysis is supported by the following examples.

## **5** Illustrative examples

In order to validate the above theoretical framework, some examples are presented in simulation and analyzed. The multi-unicycle kinematics is taken in view of the multi- $\Sigma_1$  system. Further as multi- $\Sigma_2$  examples, we propose to take a multi-second-order dynamics as system with linear drift term and multiple pendulums integrating nonidentical nonlinear drift terms. The cited examples are expected to achieve finitetime average consensus. At the second stage of the given numerical simulations, the networked dynamical systems stability is handled by tests on multiunicycle. For consensus and stability objectives, the undirected fixed networked topology (binary graph) is shown by Fig.1



Figure 1:  $\mathcal{G}$  for a system with 4 agents.

# 5.1 The multi-system finite-time consensus results

Three illustrative examples are considered here where the multi-unicycle that represents the networked systems modeled by (2), a multi-system based on second order dynamic which imply a networked multi-model as in (13), and a multi-pendulum example as in (3). Each associated protocol is deduced from (5).

### a) Average consensus in multi-unicycle

Consider N wheeled mobile robots (unicycles) where the *i*th nonholonomic kinematic model is as:

$$\begin{pmatrix} \dot{x}_i \\ \dot{y}_i \\ \dot{\theta}_i \end{pmatrix} = \begin{pmatrix} \cos(\theta_i) & 0 \\ \sin(\theta_i) & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} \dot{u}_i \\ \dot{w}_i \end{pmatrix} \qquad i = 1 \dots N$$
(30)

where  $(x_i, y_i, \theta_i)$  denotes the position and the orientation in a an inertial frame. The inputs  $u_i$  and  $w_i$  are the linear and angular velocities, respectively.

Let 
$$B = \begin{pmatrix} \cos(\theta_i) & 0\\ \sin(\theta_i) & 0\\ 0 & 1 \end{pmatrix}$$
 and  
 $C = \begin{pmatrix} \cos(\theta_i) & \sin(\theta_i & 0\\ -\sin(\theta_i) & \cos(\theta_i & 0) \end{pmatrix}$  Based on Proposition 1, the finite-time average consensus problem

can be achieved through the following protocol

$$u_i = -\sum_{j=1}^N a_{ij}\phi_\alpha(x_i - x_j)\cos(\theta_i) - \sum_{j=1}^N a_{ij}\phi_\alpha(y_i - y_j)\sin(\theta_i)$$
(31)

$$w_i = \sum_{j=1}^N a_{ij}\phi_\alpha(x_i - x_j)\sin(\theta_i) - \sum_{j=1}^N a_{ij}\phi_\alpha(y_i - y_j)\cos(\theta_i)$$
(32)

where  $\varphi_{\alpha}$  is defined in section 2 and  $a_{ij}$  are associated to the graph in Fig. 1. The simulation results are limited to N = 4 that integrate the following initial conditions

$$(x_1, y_1, \theta_1)(t = 0) = (14, 2, \pi)$$
$$(x_2, y_2, \theta_2)(t = 0) = (-4, 2, -\frac{\pi}{2})$$
$$(x_3, y_3, \theta_3)(t = 0) = (10, 8, \frac{\pi}{2})$$
$$(x_4, y_4, \theta_4)(t = 0) = (-10, -8, 0)$$



Figure 2: Average consensus of position  $x_i$  for 4 unicycles as multi- $\Sigma_1$ 

The numerical simulations are performed using (30) and protocols (31)-(32). The results of figures Fig. 2-3 evolve according to the developed theoretical results of multi- $\Sigma_1$ . The common value is also the average of the unicycles initial conditions. The

 $||(x_i, y_i) - (ave(x_i(0)), ave(y_i(0)))||$  converges in finite-time to zero as show in figure Fig.4.

#### b) Average consensus in multi-second-order dynamicss

A commonly used example in the literature is an



Figure 3: Average consensus of position  $y_i$  for 4 unicycles as multi- $\Sigma_1$ 



Figure 4: Convergence of  $||(x_i, y_i) - (ave(x_i) - ave(x_i))||$  $ave(y_i)) \parallel$ 

agent with a second-order dynamic (we can see [18])

$$\dot{x}_i = -v_i \dot{v}_i = -u_i \quad i = 1, \dots, N$$
 (33)

where  $x_i, v_i \in R$  are the states and  $u_i \in R$ is the control input. The dynamic (33) takes the form given by (13) with  $x^i = \begin{pmatrix} x_i \\ v_i \end{pmatrix}$ ,  $f^i(x^i) =$  $\begin{pmatrix} 1\\ 0 \end{pmatrix} x^i \text{ and } B = \begin{pmatrix} 0\\ 1 \end{pmatrix}$ 

For the protocol (5) we take  $C = (1 \ 1)$ . From Proposition 2 results, protocols that achieve finite time average consensus are such that

$$u_i = -\sum_{j=1}^N a_{ij}(\varphi_\alpha(x_i - x_j) + \varphi_\alpha(v_i - v_j)) \quad (34)$$

Let us take N = 4. The control parameter is taken  $\alpha = 0.5$ , and each agent initial vector of states is

as

$$(x_1, x_1, x_3, x_4)(t = 0) = (5, 10, 1, -5)$$
 (meter)

and

 $(v_1, v_1, v_3, v_4)(t = 0) = (2, -1, 8, -4)$  (meter/second)

For i = 1, ..., 4,  $x_i$  (Fig.5) and  $v_i$  (Fig.6) consent an average trajectory.



Figure 5: A reached average trajectory in positions by 4 second-order dynamics.



Figure 6: A reached average trajectory in velocities by 4 second-order dynamics.

**Remark 2** Other processes can be studied, and where the average is an agreement value of states like a common temperature of sensors where fluctuations of data is important. The energy consumption is also an important factor for stability of electric generators in networks. As example, for a multi-second-order dynamics, the kinetic energies consent an average, and this is shown by figure Fig.7.

### c) Average consensus in multi-pendulum dynamics



Figure 7: The average of kinetic energies like consensus for 4 second-order dynamics.

Consider a set of N pendulum with the following model

$$\ddot{\theta}_i = -\frac{g}{l_i}\sin(\theta_i) - -\frac{\psi_i}{m_i l_i}\dot{\theta}_i + u_i$$
(35)

where  $m_i, g_i, l_i$  and  $\psi_i$  are positive constants. For this system the drift term issues from the first order differential form (see (3)) is

$$f^{i}(\theta^{i}, \dot{\theta}_{i}) = \begin{pmatrix} \dot{\theta}_{i} \\ -\frac{g}{l_{i}}\sin(\theta_{i}) - -\frac{\psi_{i}}{m_{i}l_{i}}\dot{\theta}_{i} \end{pmatrix}$$

we can easily check the convexity condition for the drift term  $f^i$ . Following to the subsequent theoretical analysis (see Proposition 3), taking  $C = (1 \ 1)$ , a protocol that solves the finite-time average consensus for multi-pendulum is as

$$u_i = -\sum_{j=1}^N a_{ij}(\varphi_\alpha(\theta_i - \theta_j) + \varphi_\alpha(\dot{\theta}_i - \dot{\theta}_j))$$
(36)

This set of N = 4 pendulums is analyzed. As heterogenous multi-system, the 4 pendulum parameters aren't similar. Thus,  $m_1 = 1, m_2 = 2, m_3 = 3$  and  $m_4 = 4$  (Kg). The standard gravity vector is  $g = 9.8(m.s^{-2})$ , the lengths  $l_i = 1$  (m) and the coefficient  $\psi_i = 0.1(Kg.m^2.s^{-1})$ . Initial conditions are such that  $\theta_i = (-0.8, 0.4, 1, 2, 1.6)$  (rad) and  $\dot{\theta}_i = (0, 0, 0, 0)(rad.s^{-1})$ . Clearly from figures in Fig. 8-9, the synchronization toward the average trajectory of 4 pendulums in angular positions and velocities are obtained. It is important to note that the average is time-varying and the multi-system of pendulums is heterogeneous with respect to the proposed physical parameters. This confirm the theoretical results of Proposition 3.



Figure 8: The average of kinetic energies like consensus for 4 second-order dynamics.



Figure 9: The time-varying average of angular velocities consent by 4 pendulums.

# 5.2 The multi-system finite -time stability results

We consider a multi-unicycle which represents the networked system modeled by (2) (driftless). The associated protocol is deduced from (6) and the graph is in Fig.1. From Proposition 4, the finite-time stability problem is achieved for the control matrix

 $C = \begin{pmatrix} \cos(\theta_i) & \sin(\theta_i & 0) \\ -\sin(\theta_i) & \cos(\theta_i & 0) \end{pmatrix}$  that leads to the

stabilizing control-inputs

$$u_i = -\sum_{j=1}^N a_{ij}\phi_\alpha(x_i - x_j)\cos(\theta_i) - \sum_{j=1}^N a_{ij}\phi_\alpha(y_i - y_j)\sin(\theta_i)$$
(37)

$$w_i = \sum_{j=1}^N a_{ij} \phi_\alpha(x_i - x_j) \sin(\theta_i) - \sum_{j=1}^N a_{ij} \phi_\alpha(y_i - y_j) \cos(\theta_i) \mathbf{6}$$
(38)

where  $\varphi_{\alpha}$  is defined in section 2 and  $a_{ij}$  are associated to the graph in Fig. 1. Taking N = 4, the initial conditions are as T

$$(x_1, y_1, \theta_1)(t = 0) = (4, 2, \frac{\pi}{4})$$
  

$$x_2, y_2, \theta_2)(t = 0) = (12, -10, -\frac{\pi}{2})$$
  

$$(x_3, y_3, \theta_3)(t = 0) = (10, -8, \frac{2\pi}{3})$$
  

$$(x_4, y_4, \theta_4)(t = 0) = (-10, -14, \pi)$$

(

The results of stabilization are sketched in figures Fig.10-11 and the stabilizing protocols are given by figures Fig.12-13 which confirm the stability of each unicycle at the origin with continuous control feedback.



Figure 10: Finite-time stability of  $x_i$  as positions of 4 unicycles



Figure 11: Finite-time stability of  $y_i$  as positions of 4 unicycles

## Conclusion

For networked dynamic systems affine in the control vector, two protocols are proposed and theoret-



Figure 12: Stabilizing inputs  $u_i$  of 4 unicycles



Figure 13: Stabilizing inputs  $w_i$  of 4 unicycles

ically analyzed with respect to two types of nonlinear dynamic models. For a nonlinear driftless multisystem, necessary conditions on the control matrix are derived that assert finite-time average consensus toward a predefined agreement value, obtained from the multi-system initial conditions. However, for multisystem integrating drift terms, sufficient conditions on the drift term are discussed, and when they associated to the protocol solve a finite-time average consensus where as a result an average trajectory is followed by the group. Further, our stability results in networked dynamic systems overcome the individual stability analysis of each system where some obstructions for the agent's stability at the origin occur. It is well known that an unicycle doesn't verify the Brockett's necessary condition and the stabilization at the origin isn't possible with feedbacks that depend only on states. Here, due to the interconnection, the multi-unicycle stability result implies the stability of each unicycle with smooth and bounded controlinputs. The results of the paper can be extended using a directed graph while one may address the problem of consensus and stability for heterogenous systems based on the two fundamental dynamic models.

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