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Consensuability Conditions of Multi Agent Systems with Varying Interconnection Topology and Different Kinds of Node Dynamics

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1. Introduction

Many systems in nature and of practical interest can be modeled as large collections of interacting subsystems. Such systems are referred as "Multi Agent Systems" (briefly MASs) and some examples include electrical power distribution networks (P. Kundur, 1994), communication (F. Paganini, 2001), and collections of vehicles traveling in formation (J.K. Hedrick et al., 1990). Several practical issues concern the design of decentralized controllers and the stability analysis of MASs in the presence of uncertainties in the subsystem interconnection topology (i.e. due in practical applications to failures of transmission lines). The analysis and control of collections of interconnected systems have been widely studied in the literature. Early work on stability analysis and decentralized control of large-scale interconnected systems is found in (D. Limebeer & Y.S. Hung, 1983; A. Michel & R. Miller, 1977; P.J. Moylan & D.J. Hill, 1978; Siljak, 1978; J.C. Willems, 1976). Some of the more widely notable stability criteria are based on the passivity conditions (M. Vidyasagar, 1977) and on the well-known notion of connective stability introduced in (Siljak, 1978).

More recently, MASs have appeared broadly in several applications including formation flight, sensor networks, swarms, collective behavior of flocks (Savkin, 2004; C.C. Cheaha et al., 2009; W. Ren, 2009) motivating the recent significant attention of the scientific community to distributed control and consensus problems (i.e. (R.O. Saber & R. Murray, 2004; Z. Lin et al., 2004; V. Blondel et al., 2005; J. N. Tsitsiklis et al., 1986)). One common feature of the consensus algorithm is to allow every agent automatically converge to a common *consensus state* using only local information received from its neighboring agents. "Consensuability" of MASs is a fundamental problem concerning with the existence conditions of the consensus state and it is of great importance in both theoretical and practical features of cooperative protocol (i.e. flocking, rendezvous problem, robot coordination). Results about consensuability of MASs are related to first and second order systems and are based on the assumption of jointly-connected interaction graphs (i.e. in (R.O. Saber & R. Murray, 2004; J. N. Tsitsiklis et al., 1986)). Extension to more general linear MASs whose agents are described by LTI (Linear Time Invariant) systems can be found in (Tuna, 2008) where the closed-loop MASs were shown to be asymptotic consensus stable if the topology had a spanning tree. In (L. Scardovi & R. Sepulchre, 2009) it is investigated the synchronization of a

network of identical linear state-space models under a possibly time-varying and directed interconnection structure. Many investigations are carried out when the dynamic structure is fixed and the communication topology is time varying (i.e. in (R.O. Saber & R. Murray, 2004; W. Ren & R. W. Beard, 2005; Ya Zhanga & Yu-Ping Tian, 2009)). One of main appealing field of research is the investigation of the MASs consensusability under both the dynamic agent structure and communication topology variations. In particular, it is worth analyzing the joint impact of the agent dynamic and the communication topology on the MASs consensusability. The aim of the chapter is to give consensusability conditions of LTI MASs as function of the agent dynamic structure, communication topology and coupling strength parameters. The theoretical results are derived by transferring the consensusability problem into the robust stability analysis of LTI-MASs. Differently from the existing works, here the consensusability conditions are given in terms of the adjacency matrix rather than Laplacian matrix. Moreover, it is shown that the interplay among consensusability, node dynamic and topology must be taken into account for MASs stabilization: specifically, consensusability of MASs is assessed for all topologies, dynamic and coupling strength satisfying a pre-specified bound. From the practical point of view the consensusability conditions can be used for both the analysis and planning of MASs protocols to guarantee robust stability for a wide range of possible interconnection topologies, coupling strength and node dynamics. Also, the number of subsystems affecting the overall system stability is taken into account as it is analyzed the robustness of multi agent systems if the number of subsystems changes. Finally, simulation examples are given to illustrate the theoretical analysis.

2. Problem statement

We consider a network composed of linear systems interconnected by a specific topological structure. The dynamical system at each node is of m -th order and described by the matrices (A, B, C) . Let $G(V, E, U)$ be a directed weighted graph (digraph) with the set of nodes $V = 1..n$, set of edges $E \subseteq n \times n$, and the associated weighted adjacency matrix $U = \{u_{ij}\}$ with $u_{ij} > 0$ if there is a directed edge of weight u_{ij} from vertex j (node parent) into vertex i (node child). The linear systems are interconnected by a directed weighted graph $G(V, E, U)$. Each node dynamical is described by:

$$\begin{aligned} \dot{x}_i(t) &= Ax_i(t) + Bv_i(t) \\ y_i(t) &= Cx_i(t) \end{aligned} \quad (1)$$

with $v_i(t)$ is the input to the i -th node of the form

$$v_i(t) = \sum_{j=1}^n u_{ij}y_j(t). \quad (2)$$

In this way, each node dynamic is influenced by the sum of its neighbors' outputs. This yields to the MAS network equation:

$$\dot{x}_i(t) = Ax_i(t) + \sum_{j=1}^n u_{ij}BCx_j(t) \quad (3)$$

with $1 \leq i \leq n$, and its compact form:

$$\dot{x}(t) = A_\phi x(t) \quad (4)$$

with $A_g = (I_n \otimes A) + (U \otimes BC)$, with \otimes denotes the matrix Kronecker product. Notice that the above equation can be associated to the main model used in the literature for describing the synchronization phenomena, energy distribution, tanks network (e.g. in (R. Cogill & S. Lall, 2004)). Moreover the system at each node can be MIMO or SISO type, and the matrix product BC takes into account the coupling strength and the coupling interaction among the state system variables. Observing the MAS model (3) we point out as the overall network dynamic is affected by the node system dynamic matrix A , the coupling matrix BC , and by the adjacency matrix U of the topological structure.

Consider a network with n agents whose topology information exchange is described by a graph $G(V, E, U)$ and let x_i the state of agent-node i -th, consensus corresponds to the network condition such that the state of the agents as a whole asymptotically converges to an equilibrium state with identical elements (i.e. $x_i = x_j$ for all $(i, j) \in n \times n$). The common value \bar{x} is named *consensus value*. Consensusability of MASs is a fundamental problem concerning with the conditions for assessing network consensus equilibrium. Under the assumption of the existence of a network equilibrium, then consensusability deals with the research of analytical conditions such that the network equilibrium corresponds to a consensus state. In this way, without loss of generality, the consensusability problem can be reduced to the problem of assessing stabilization conditions of the MAS network (3) with respect to the 0 equilibrium point (i.e. $x_i = x_j = 0$ for all $(i, j) \in n \times n$).

Hence, we are interested in solving the following problem:

Problem Given a multi agent network described by (3), to determinate the MAS consensusability conditions as function of node dynamic, topology and coupling strength.

Specifically, consensusability of MASs is assessed for all topologies, dynamic and coupling strength satisfying a pre-specified bound.

In the follows we will present analytical conditions for solving the above Problem.

3. Conditions for MASs consensusability

Before of presenting the MASs consensusability conditions of (3), we have to recast the eigenvalues set $\sigma(A_g)$ of MAS network dynamic matrix A_g .

Lemma 1 Let $\sigma(U)=\{\mu_i\}$ the eigenvalues set of the adjacency matrix U , $\sigma(A_g)$ the eigenvalues set of the MAS dynamical matrix A_g , then results: $\sigma(A_g) = \cup_i \sigma(A + \mu_i BC)$ for all $1 \leq i \leq n$.

Proof Let J the Jordan canonical form of U , then it exists a similarity matrix S so that $J = S^{-1}US$. Hence $S \otimes I_n$ is a similarity matrix for the matrices $I_m \otimes A + U \otimes BC$ and $I_m \otimes A + J \otimes BC$. From the Kronecker product (Horn R.A. & Johnson C.R., 1995) results:

$$\begin{aligned} (S \otimes I_n)^{-1}(I_m \otimes A + U \otimes BC)(S \otimes I_n) &= \\ (S^{-1} \otimes I_n)(I_m \otimes A + U \otimes BC)(S \otimes I_n) &= \\ (I_m \otimes A) + (S^{-1}US \otimes BC) &= (I_m \otimes A) + (J \otimes BC) \end{aligned}$$

with J being an upper triangular matrix with $I_m \otimes A + J \otimes BC$ as upper triangular block matrix. Hence the eigenvalues of the matrix $I_m \otimes A + J \otimes BC$ are the union of the eigenvalues of the block matrix on the diagonal.

From the above Lemma 1, the eigenvalues of the MAS dynamic matrix A_g are explicitly function of those of the matrix $A + \mu_i BC$, for all i . So we can decouple the effects of topology structure (by μ_i), the coupling strength BC and node dynamic A on the overall stability of the MAS. This can be used for giving stability MAS condition as function of topology structure, node dynamic and coupling strength as shown by the following Theorem 1:

Theorem 1 Let the MAS composed of n identical MIMO system of order m -th and interconnected by the digraph $G = (V, E, U)$ with adjacency matrix U , with eigenvalues $\mu_1 \leq \mu_2 \leq \dots \mu_n$. If the node dynamic matrix $A = \{a_{ij}\}$ and the coupling matrix $BC = \{c_{ij}\}$ fulfill the conditions:

$$\begin{aligned} a_{ii} + \mu_k c_{ii} &\leq 0 \\ |a_{ii} + \mu_k c_{ii}| &\geq \sum_{j \neq i} |a_{ij} + \mu_k c_{ij}| \end{aligned} \quad (5)$$

$\forall i = 1, 2, \dots, m$ and $\forall k = 1, 2, \dots, n$, then the MAS (3) is stable.

Proof If the conditions (5) hold, then all eigenvalues of the matrix

$$A + \mu_k BC = \begin{pmatrix} a_{11} + \mu_k c_{11} & a_{12} + \mu_k c_{12} & \dots & a_{1m} + \mu_k c_{1n} \\ a_{21} + \mu_k c_{21} & a_{22} + \mu_k c_{22} & \dots & a_{2m} + \mu_k c_{2m} \\ \vdots & \vdots & \vdots & \ddots \\ a_{m1} + \mu_k c_{m1} & a_{m2} + \mu_k c_{m2} & \dots & a_{mm} + \mu_k c_{mm} \end{pmatrix},$$

$\forall k = 1, 2, \dots, n$, are located in a convex set in the left complex half plane as result by the application of the Gershgorin's circle theorem (Horn R.A. & Johnson C.R., 1995). Hence, by Lemma 1, the MAS is stable.

The previous Theorem 1 easily yields to the following corollaries.

Corollary 1 Let the MAS composed of n identical MIMO system of order 2 and interconnected by the digraph $G = (V, E, U)$, with adjacency matrix U with eigenvalues $\mu_1 \leq \mu_2 \leq \dots \mu_n$. If the node dynamic matrix $A = \{a_{ij}\}$ and the coupling matrix $BC = \{c_{ij}\}$ with $c_{ij} \geq 0, i, j = 1, 2$, fulfill the conditions:

$$\begin{aligned} a_{ij} &\geq -c_{ij}\mu_1 \\ a_{ii} &\leq -a_{ij} - (c_{ii} + c_{ij}) \cdot \mu_n \end{aligned} \quad (6)$$

or

$$\begin{aligned} a_{ii} &\leq -c_{ii}\mu_n \\ a_{ij} &\leq -c_{ij}\mu_n \\ a_{ii} &\leq a_{ij} + (c_{ij} - c_{ii}) \cdot \mu_1, \end{aligned} \quad (7)$$

$i, j = 1, 2$, then the MAS (3) is stable.

Because the adjacency matrix U of a graph has both positive and negative eigenvalues, the conditions (6) and (7) implicitly imply the assumption that the single system at the node is stable. In this way, as expected, we derive that it is not possible to stabilize a network of instable systems by acting only on the topological structure. Given a specified node dynamic, coupling strength and bound on the adjacency matrix U , by conditions (6) and (7) we can assess MAS stability. Moreover, the MAS robustness with respect to varying switching topology can be dealt by considering the span of the eigenvalue of the admissible structure topologies. As we will show in the follows, it is possible easily to evaluate the eigenvalues of A_g , given the eigenvalues of U in some simple and representative cases of interest.

Corollary 2 Let the MAS composed of n identical MIMO system of order 1 and interconnected by the digraph $G = (V, E, U)$, with adjacency matrix U with eigenvalues $\mu_1 \leq \mu_2 \leq \dots \mu_n$. If the node dynamic matrix $A = a$ and the coupling matrix $BC = c$ fulfill the conditions:

$$a \leq -c \cdot \mu_n \quad \text{if } c \geq 0 \tag{8}$$

$$a \leq -c \cdot \mu_1 \quad \text{if } c < 0, \tag{9}$$

then the MAS (3) is stable.

The Corollary 2 reduces the analytical result of Theorem 1 to the case of the consensus of integrator (R.O. Saber & R. Murray, 2004) with coupling gain c . Smaller c , higher is the degree of robustness of the network to the slower node dynamic. In the opposite, higher c reduces the stability margin of the MAS. Finally, for a fixed dynamic at the node, the maximum admissible coupling strength c depends on the maximum and minimum eigenvalues of the adjacency matrix:

$$c \leq -\frac{a}{\mu_n} \quad \text{if } c \geq 0 \tag{10}$$

$$c \geq -\frac{a}{\mu_1} \quad \text{if } c < 0. \tag{11}$$

Corollary 3 Let the MAS of n identical MIMO system of the m -th order, interconnected by the digraph $G = (V, E, U)$, with adjacency matrix U with eigenvalues $\mu_1 \leq \mu_2 \leq \dots \mu_n$. If the node dynamic matrix $A = \{a_{ij}\}$ and the coupling matrix $BC = \{c_{ij}\}$ are both upper or lower triangular matrix and fulfill the conditions:

$$a_{ii} \leq -c_{ii} \cdot \mu_n \quad \text{if } c_{ii} \geq 0 \tag{12}$$

$$a_{ii} \leq -c_{ii} \cdot \mu_1 \quad \text{if } c_{ii} < 0, \tag{13}$$

then the MAS (3) is stable.

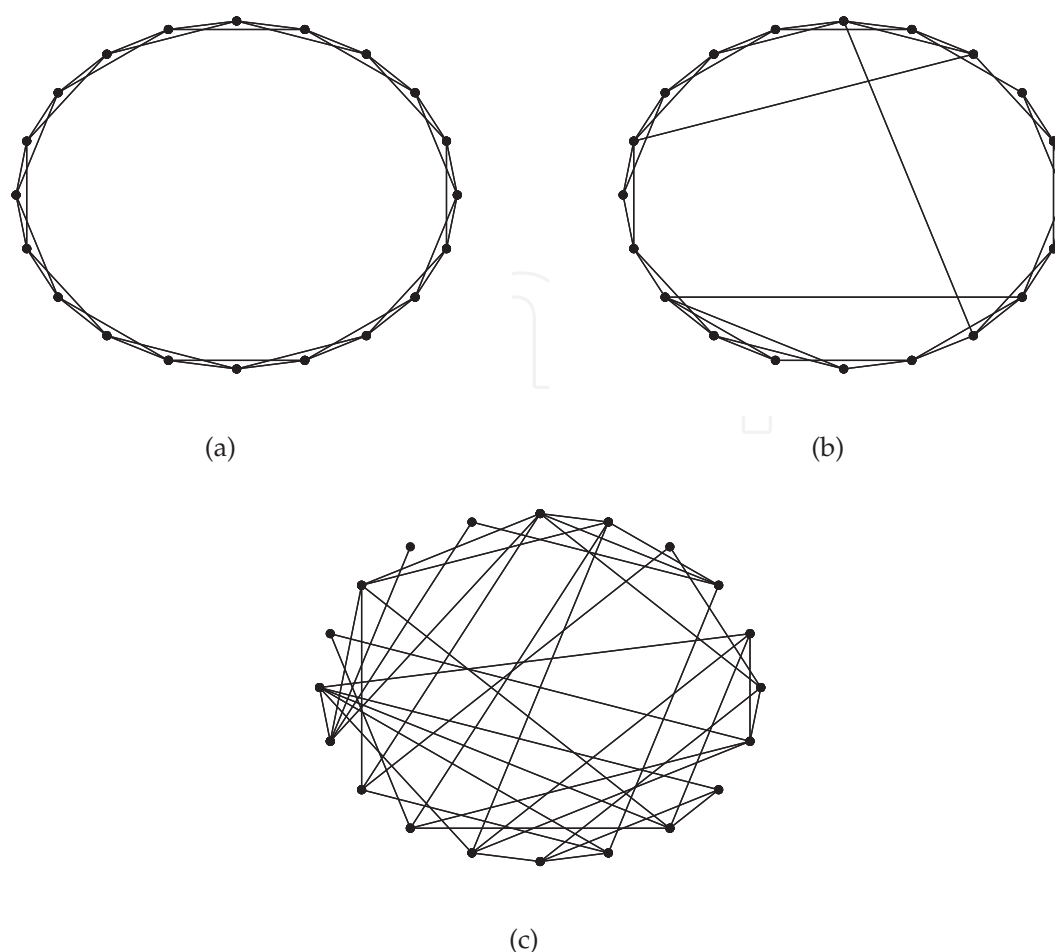


Fig. 1. Procedure of redirectioning of links in a regular network (a) with increasing probability p . As p increases the network moves from regular (a) to random (c), becoming small world (b) for a critical value of p . $n=20, k=4$

Notice that if $BC = c \cdot I_n$, (10) and (11) become:

$$c \leq -\frac{\min_i |a_{ii}|}{\mu_n} \quad \text{If } c_{ii} \geq 0 \quad (14)$$

$$c > -\frac{\min_i |a_{ii}|}{\mu_1} \quad \text{If } c_{ii} < 0 \quad (15)$$

and hence the stability of MAS is explicitly given as function of the network slowest node dynamic.

Now we would like to point out the case of undirected topology with symmetric adjacency matrix U . If we assume A and BC being symmetric, then A_g is symmetric with real eigenvalue. Moreover from the field value property (Horn R.A. & Johnson C.R., 1995), let $\sigma(A) = \{\alpha_j\}$ and $\sigma(BC) = \{v_j\}$ the eigenvalues set of A and BC , then the eigenvalues of $A + \mu_i BC$ are in the interval $[\min_j \{\alpha_j\} + \mu_i \min_j \{v_j\}, \max_j \{\alpha_j\} + \mu_i \max_j \{v_j\}]$, for every $1 \leq i \leq n, 1 \leq j \leq m$. In this way, there is a bound need to be satisfied by the topology structure, node dynamic and coupling matrix for MAS stabilization.

In the literature, the MAS consensuability results have been given in terms of Laplacian matrix properties. Here, differently, we have given bounds as function of the adjacency matrix features. Anyway we can use the results on the Laplacian eigenvalue for recasting the bounds given on the adjacency matrix. To this aim, defined the degree d_i of i -th node of an undirected graph as $\sum_j u_{ij}$, the Laplacian matrix is defined as $L = D - U$ with D is the diagonal matrix with the degree of node i -th in position i -th. Clearly L is a zero row sums matrix with non-positive off-diagonal elements. It has at least one zero eigenvalue and all nonzero eigenvalues have nonnegative real parts. So $U = D - L$ and being the minimum and maximum Laplacian eigenvalues respectively bounded by 0 and the highest node degree, we have:

Lemma 2 Let U the adjacency matrix of undirected and connected graph $G = (V, E, U)$, with eigenvalues $\mu_1 \leq \mu_2 \leq \dots \leq \mu_n$, then results:

$$\mu_1(U) \geq \min_i d_i - \min_{k,j} (\max\{d_k + d_j : (k, j) \in E(G)\}, n) \tag{16}$$

$$\mu_n(U) \leq \max_i d_i \tag{17}$$

Proof Easily follows from the Laplacian eigenvalues bound and the field value property (Horn R.A. & Johnson C.R., 1995).

4. Simulation validation

In the follows we will present a variety of simulations to validate the above theoretical results under different kinds of node dynamic and network topology variations. Specifically the MAS topology variations have been carried out by using the well known Watts-Strogatz procedure described in (Watts & S. H. Strogatz, 1998). In particular, starting from the regular network topology ($p = 0$), by increasing the probability p of rewiring the links, it is possible smoothly to change its topology into a random one ($p = 1$), with small world typically occurring at some intermediate value. In so doing neither the number of nodes nor the overall number of edges is changed. In Fig. 1 it shown the results in the case of MAS of 20 nodes with each one having $k = 4$ neighbors.

Among the simulation results we focus our attention on the maximum and minimum eigenvalues of the matrixes U (i.e. μ_n and μ_1) and A_g (i.e. λ_M and λ_m) and their bounds computed by using the results of the previous section. In particular, by Lemma 2, we convey the bounds on U eigenvalues in bounds on A_g eigenvalues suitable for the case of time varying topology structure. We assume in the simulations the matrices A and BC to be symmetric. In this way, if U eigenvalues are in $[v_1, v_2]$, let $\sigma(A) = \{\alpha_i\}$, $\sigma(BC) = \{v_i\}$, the eigenvalues of A_g will be in the interval $[\min_i \alpha_i + \min_j \{v_1 v_j, v_2 v_j\}, \max_i \alpha_i + \max_j \{v_1 v_j, v_2 v_j\}]$ for $i, j = 1, 2, \dots, n$.

Notice that, known the interval of variation $[v_1, v_2]$ of the eigenvalues set of U under switching topologies, we can recast the conditions (8), (9), (12), (13), (6), (7) and to use it for design purpose. Specifically, given the interval $[v_1, v_2]$ associated to the topology possible variations, we derive conditions on A or BC for MAS consensuability.

We consider a graph of $n = 400$ and $k = 4$. In the evolving network simulations, we started with $k = 4$ and bounded it to the order of $O(\log(n))$ for setting a sparse graph. In Tab 1 are drawn the node dynamic and coupling matrices considered in the first set of simulations.

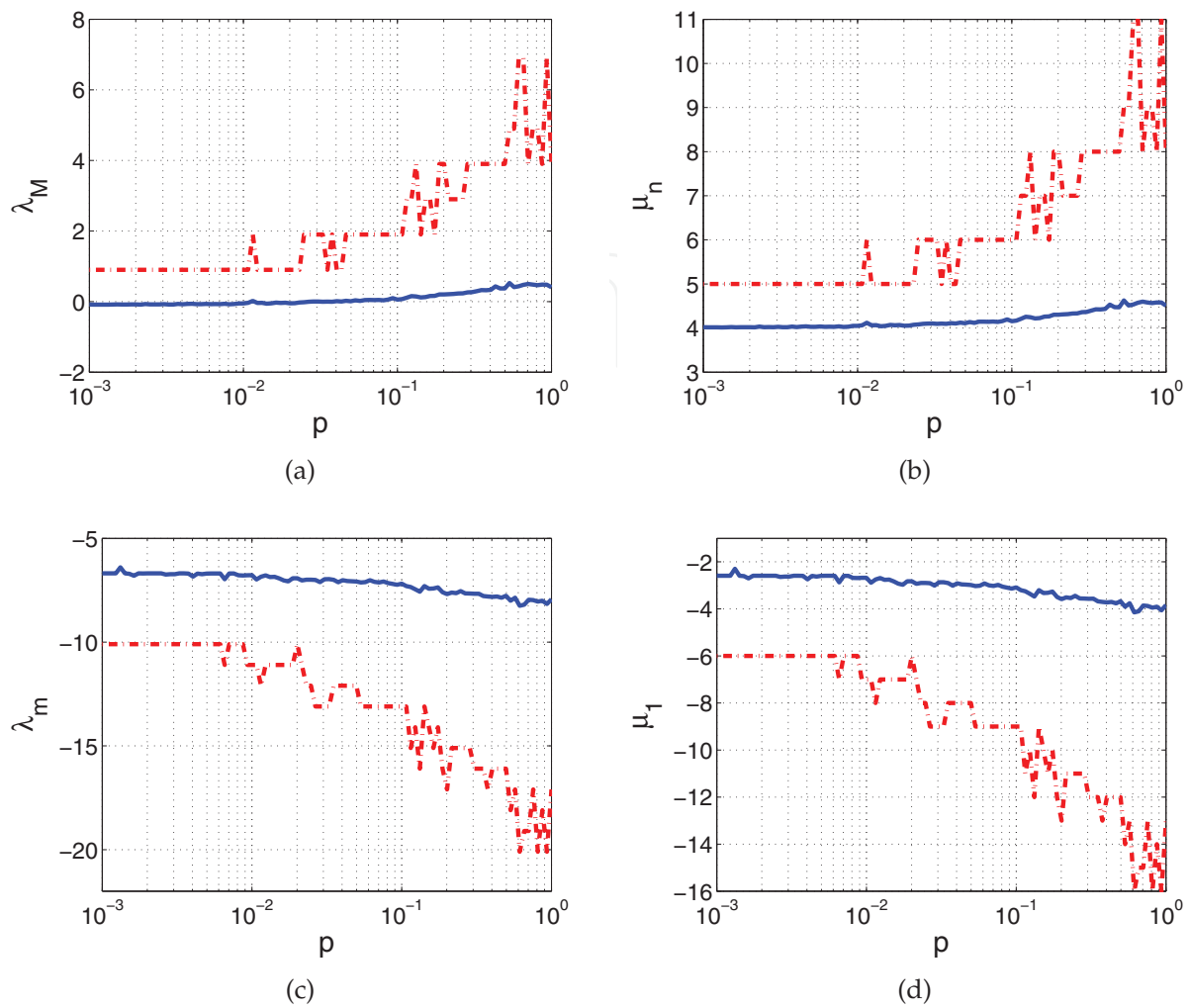


Fig. 2. Case 1. Dashed line: bound on the eigenvalue; continuous line: eigenvalues, (a) Maximum eigenvalue of A_g , (b) Maximum eigenvalue of U , (c) Minimum eigenvalue of A_g , (d) Minimum eigenvalue of U

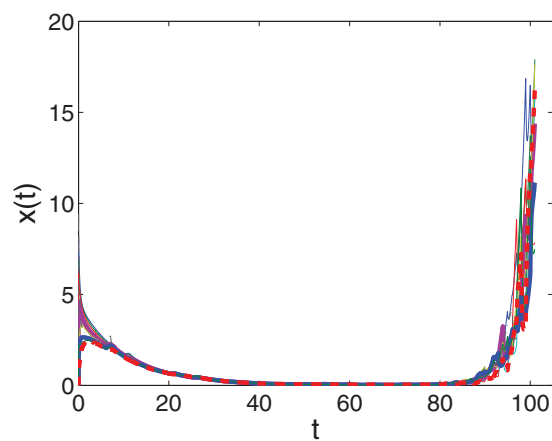


Fig. 3. Case 1: State dynamic evolution in the time

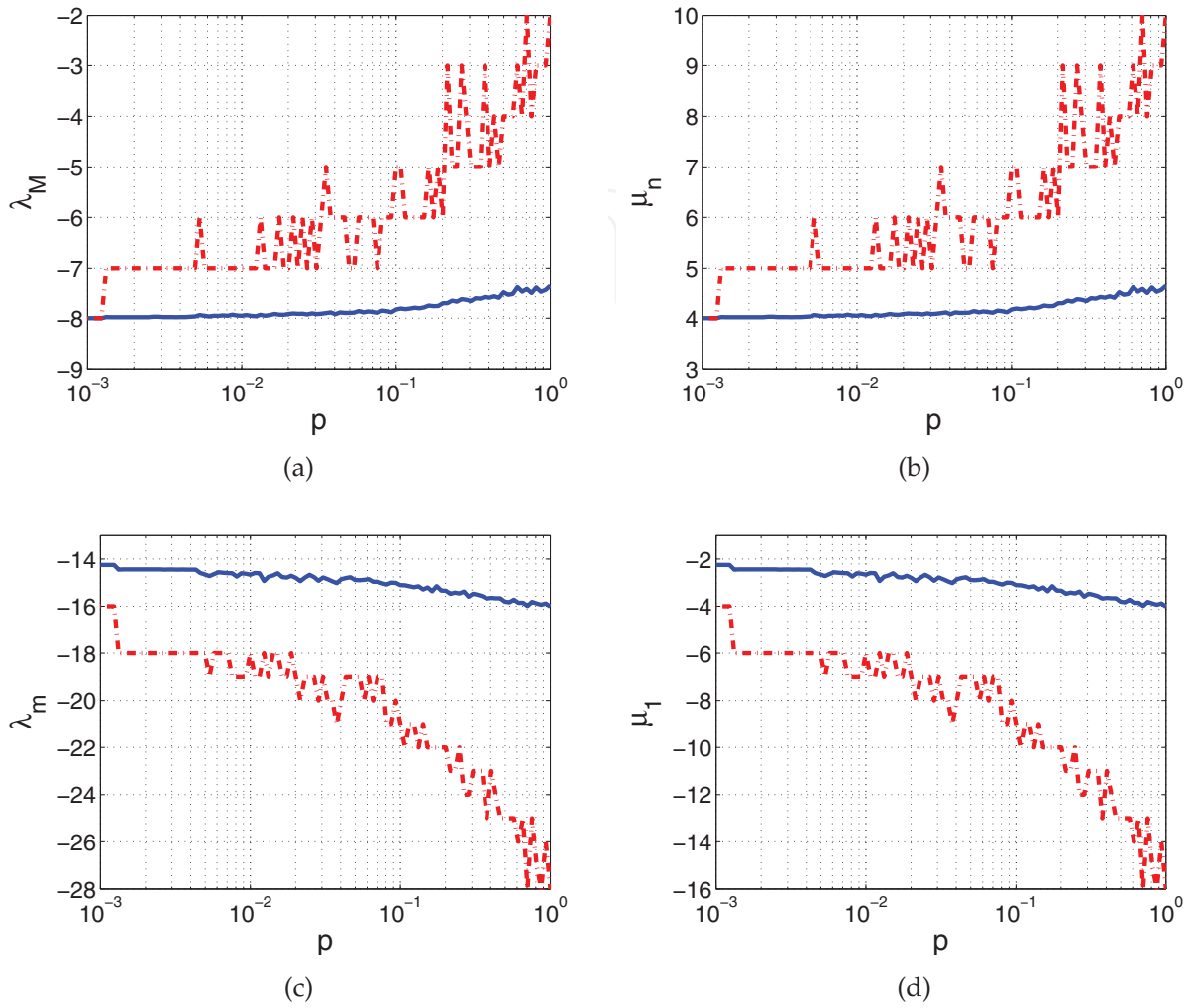


Fig. 4. Case 2. Dashed line: bound on the eigenvalue; continuous line: eigenvalue: (a) Maximum eigenvalue of A_g , (b) Maximum eigenvalue of U , (c) Minimum eigenvalue of A_g , (d) Minimum eigenvalue of U

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	A	B	C
Case 1:	-4.1	1	1
Case 2:	-12	1	1
Case 3:	-6	1	1
Case 4:	-6	2	1

Table 1. Node system matrices (A,B,C)

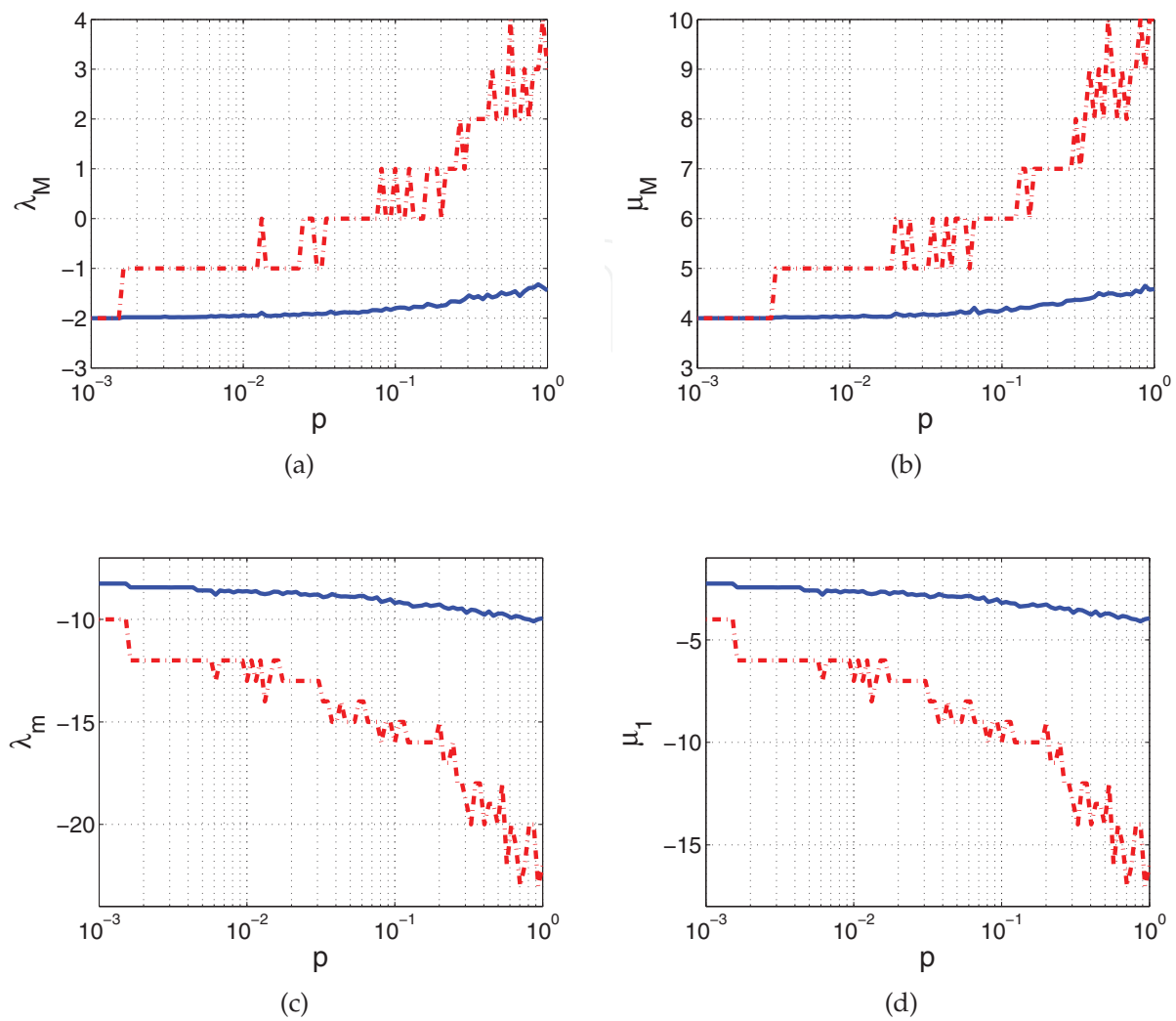


Fig. 5. Case 3. Dashed line: bound on the eigenvalue; continuous line: eigenvalue: (a) Maximum eigenvalue of A_g , (b) Maximum eigenvalue of U , (c) Minimum eigenvalue of A_g , (d) Minimum eigenvalue of U

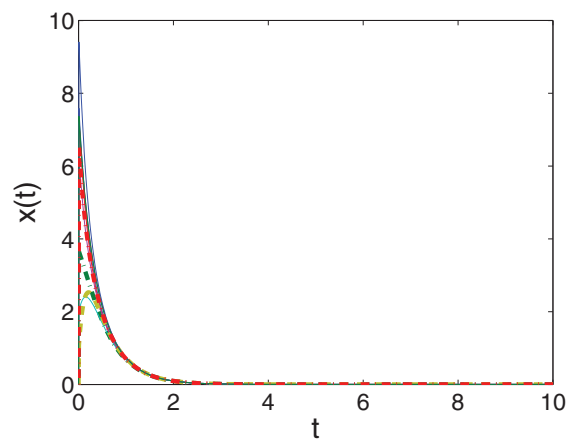


Fig. 6. Case 3: state dynamic evolution in the time

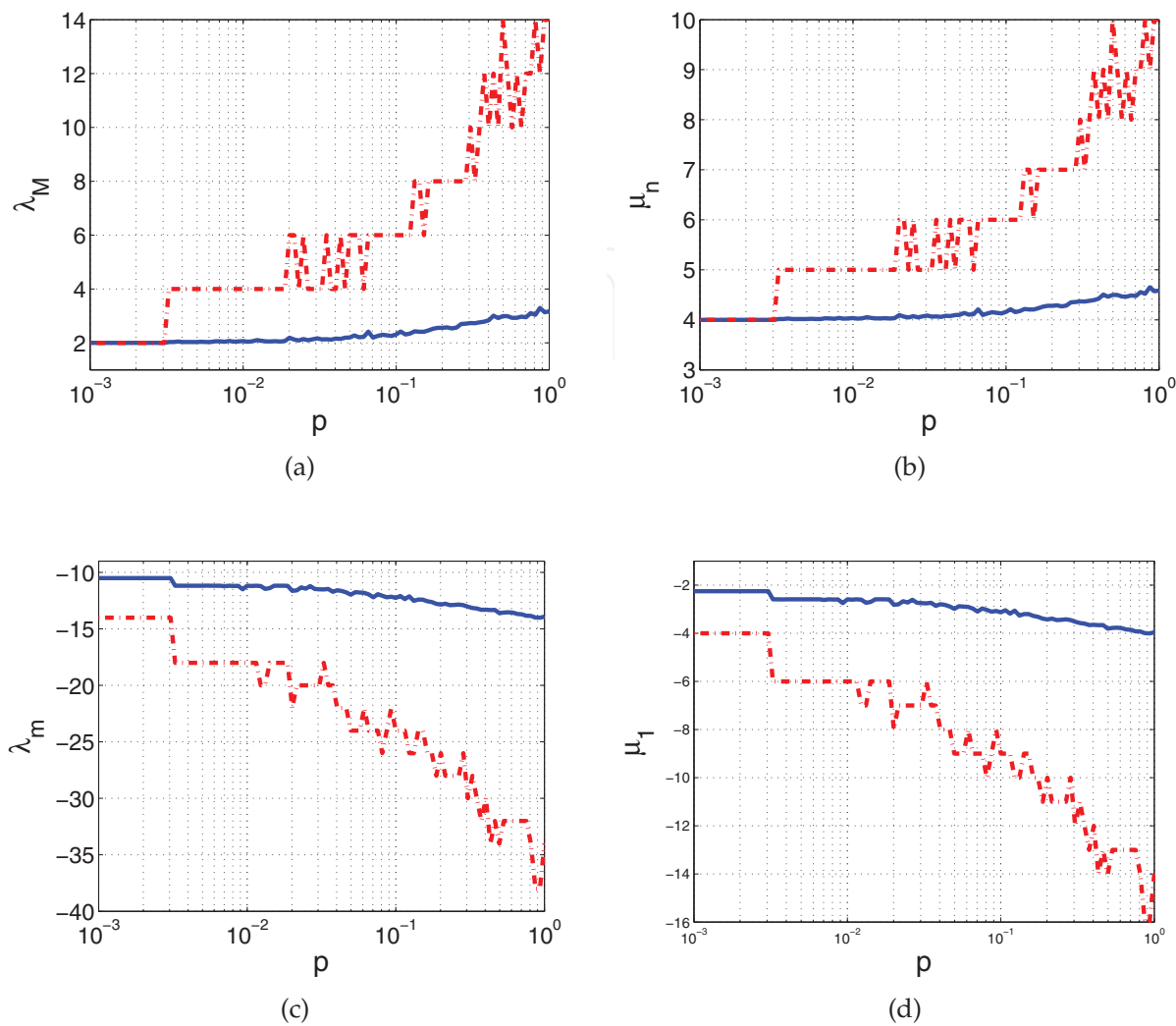


Fig. 7. Case 4. Dashed line: bound on the eigenvalues; continuous line: eigenvalues: (a) Maximum eigenvalue of A_g , (b) Maximum eigenvalue of U , (c) Minimum eigenvalue of A_g , (d) Minimum eigenvalue of U

In the case 1 (Fig 2), we note as although we start from a stable MAS network, the topology variation leads the network instability condition (namely λ_M becomes positive). In Fig. 3 it is shown the time state evolution of the firsts 10 nodes, under the switching frequency of 1 Hz. We note as the MAS converges to the consensus state till it is stable, then goes in instability condition.

In the case 2, we consider a node dynamic faster than the maximum network degree d_M of all evolving network topologies from compete to random graph. Notice that although this assures MAS consensuability as drawn in Fig. 4, it can be much conservative.

In the case 3 (Fig 5), we consider a slower node dynamic than the cases 2. The MAS is robust stable under topology variations. In Fig. 6 the state dynamic evolution is convergent and the settling time is about $4.6/|\lambda_M(A_g)|$.

Then we have varied the value for BC by doubling the B matrix value leaving unchanged the node dynamic matrix. As appears in Fig. 7, the MAS goes in instability condition pointing out

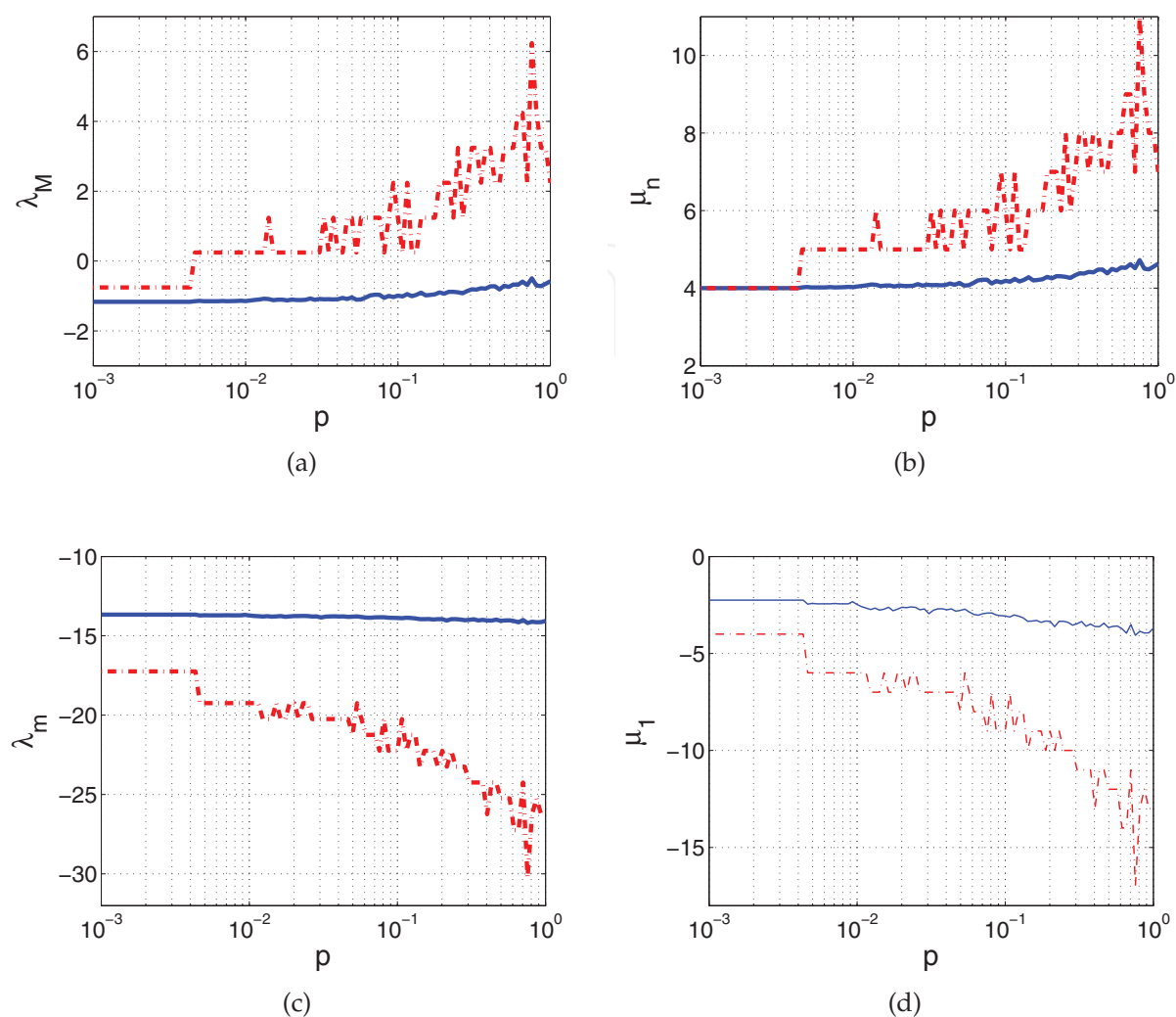


Fig. 8. Case 5. Dashed line: bound on the eigenvalue; continuous line: eigenvalue: (a) Maximum eigenvalue of A_g , (b) Maximum eigenvalue of U , (c) Minimum eigenvalue of A_g , (d) Minimum eigenvalue of U

that also the coupling strength can affect the stability (as stated by the conditions (8), (9)) and that this effect can be amplified by the network topological variations.

	A	B	C
Case 5:	$\begin{bmatrix} -6 & 3 \\ 3 & -12 \end{bmatrix}$	$\begin{bmatrix} 1 \\ 0 \end{bmatrix}$	$[1 \ 0]$
Case 6:	$\begin{bmatrix} -3 & 3 \\ 3 & -6 \end{bmatrix}$	$\begin{bmatrix} 1 \\ 0 \end{bmatrix}$	$[1 \ 0]$
Case 7:	$\begin{bmatrix} -3 & 3 \\ 3 & -6 \end{bmatrix}$	$\begin{bmatrix} 0.25 \\ 0 \end{bmatrix}$	$[1 \ 0]$

Table 2. Node system matrices (A, B, C).

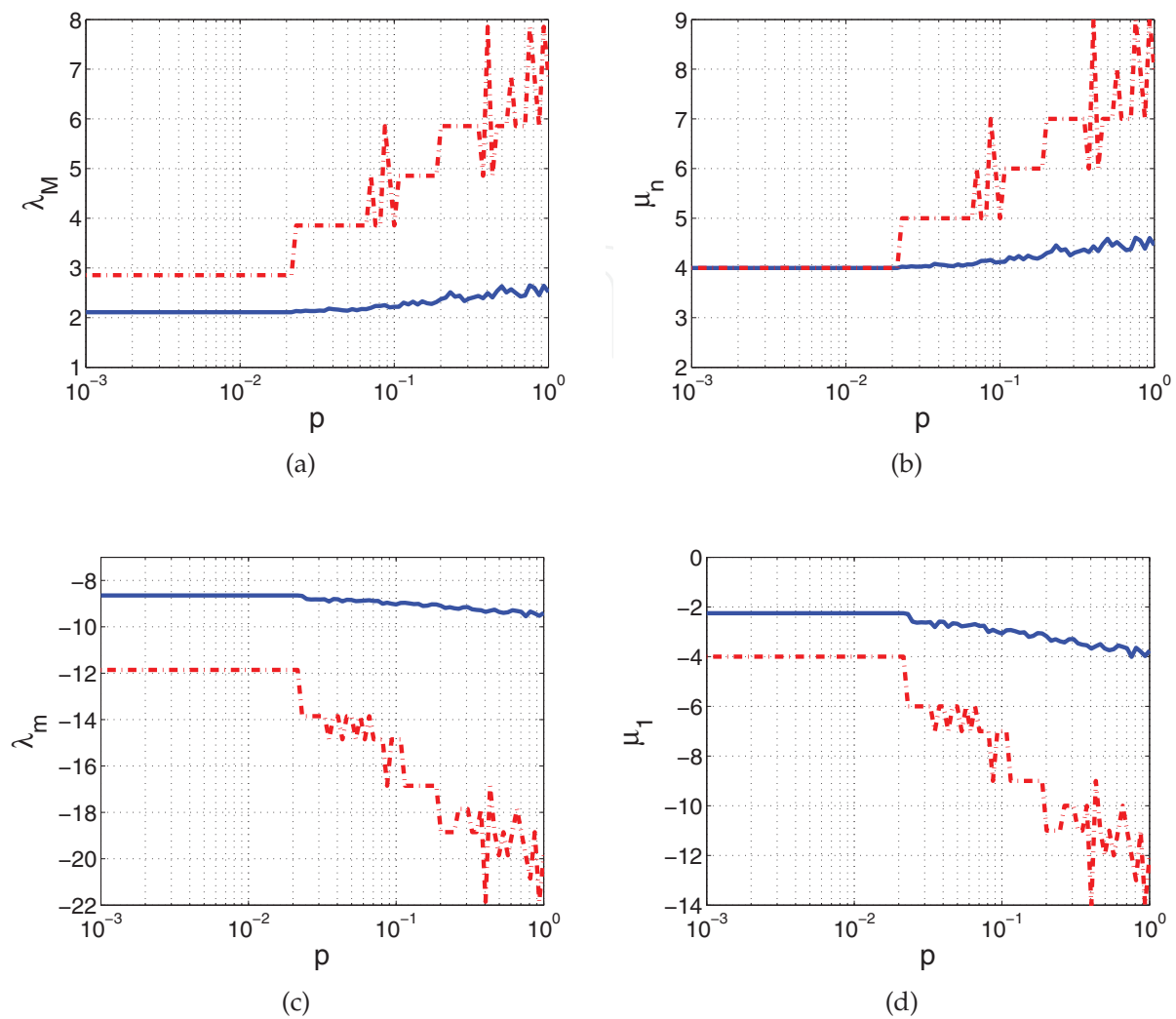


Fig. 9. Case 6. Dashed line: bound on the eigenvalue; continuous line: eigenvalue: (a) Maximum eigenvalue of A_g , (b) Maximum eigenvalue of U , (c) Minimum eigenvalue of A_g , (d) Minimum eigenvalue of U

On the other side, a reduction on BC increases the MAS stability margin. So we can tune the BC value in order to guarantee stability or desired robust stability MAS margin under a specified node dynamic and topology network variations. Indeed if BC has eigenvalues above 1, its effect is to amplify the eigenvalues of U and we need a faster node dynamic for assessing MAS stability. If BC has eigenvalues less of 1, its effect is of attenuation and the node dynamic can be slower without affecting the network stability.

Now we consider SISO system of second order at the node as shown in Tab.2. In this case the matrix BC has one zero eigenvalue being the rows linearly dependent.

In the case 5 the eigenvalues of A are $\alpha_1 = -4.76$ and $\alpha_2 = -13.23$, the eigenvalues of the coupling matrix BC are $\nu_1 = 1$ and $\nu_2 = 0$. In this case the node dynamic is sufficiently fast for guaranteeing MAS consensuability (Fig. 8). In the case 6, we reduce the node dynamic matrix A to $\alpha_1 = -1.15$ e $\alpha_2 = -7.85$. Fig. 9 shows instability condition for the MAS network. We

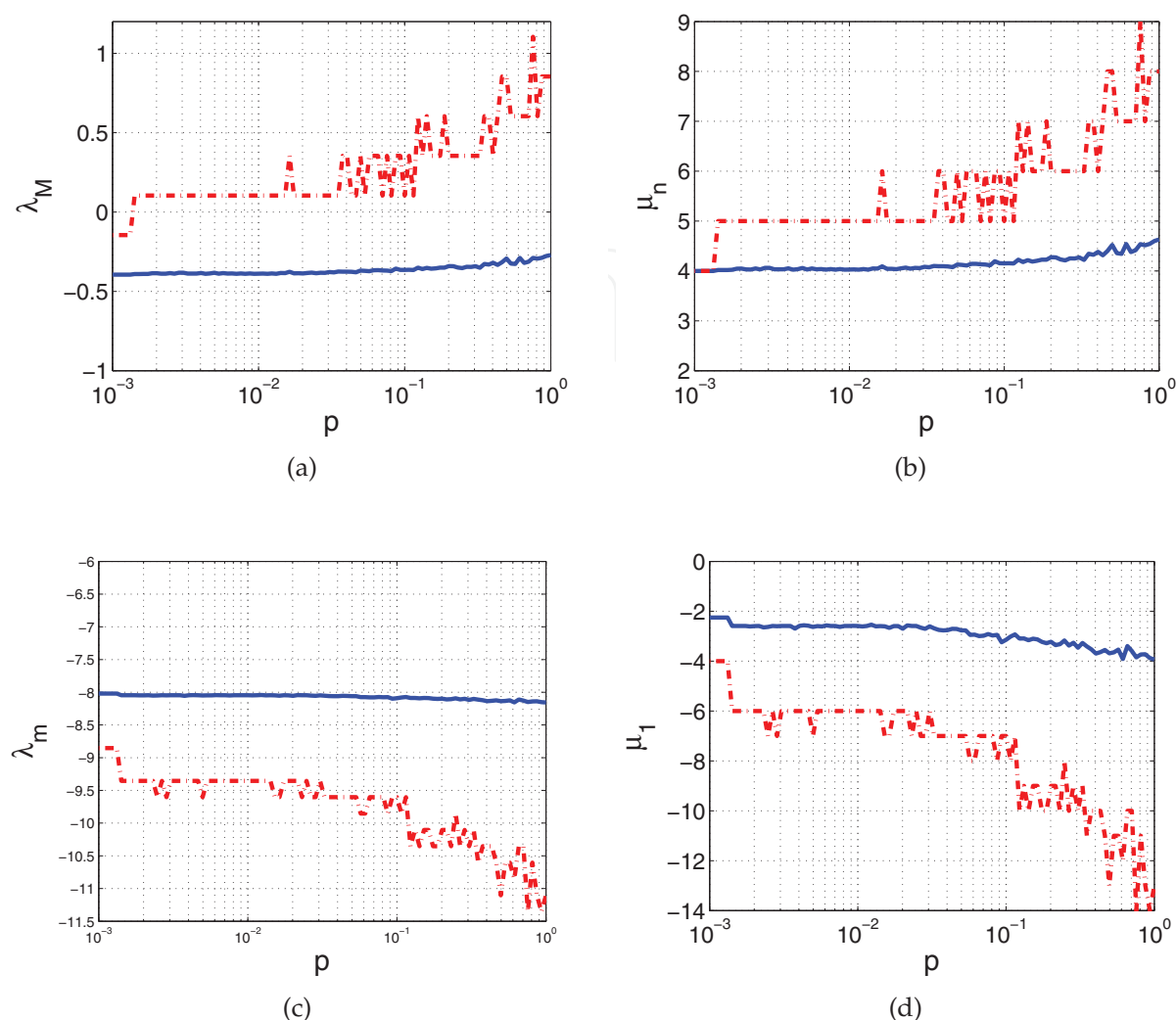


Fig. 10. Case 7. Dashed line: bound on the eigenvalue; continuous line: eigenvalue. (a) Maximum eigenvalue of A_g , (b) Maximum eigenvalue U , (c) Minimum eigenvalue of A_g , (d) Minimum eigenvalue of U

can lead the MAS in stability condition by designing the coupling matrix BC as appear by the case 7 and the associate Fig. 10.

4.1 Robustness to node fault

Now we deal with the case of node fault. We can state the following Theorem.

Theorem 2 Let A and BC symmetric matrix and $G(V, E, U)$ an undirected graph. If the MAS system described by A_g is stable, it is stable also in the presence of node faults. Moreover the MAS dynamic becomes faster after the node fault.

Proof Being the graph undirected and A and BC symmetric then A_g is symmetric. Let \tilde{A}_g the MAS dynamic matrix associated to the network after a node fault. \tilde{A}_g is obtained from A_g by eliminating the rows and columns corresponding to the nodes went down. So \tilde{A}_g is a minor of A_g and for the interlacing theorem (Horn R.A. & Johnson C.R., 1995) it has eigenvalues inside

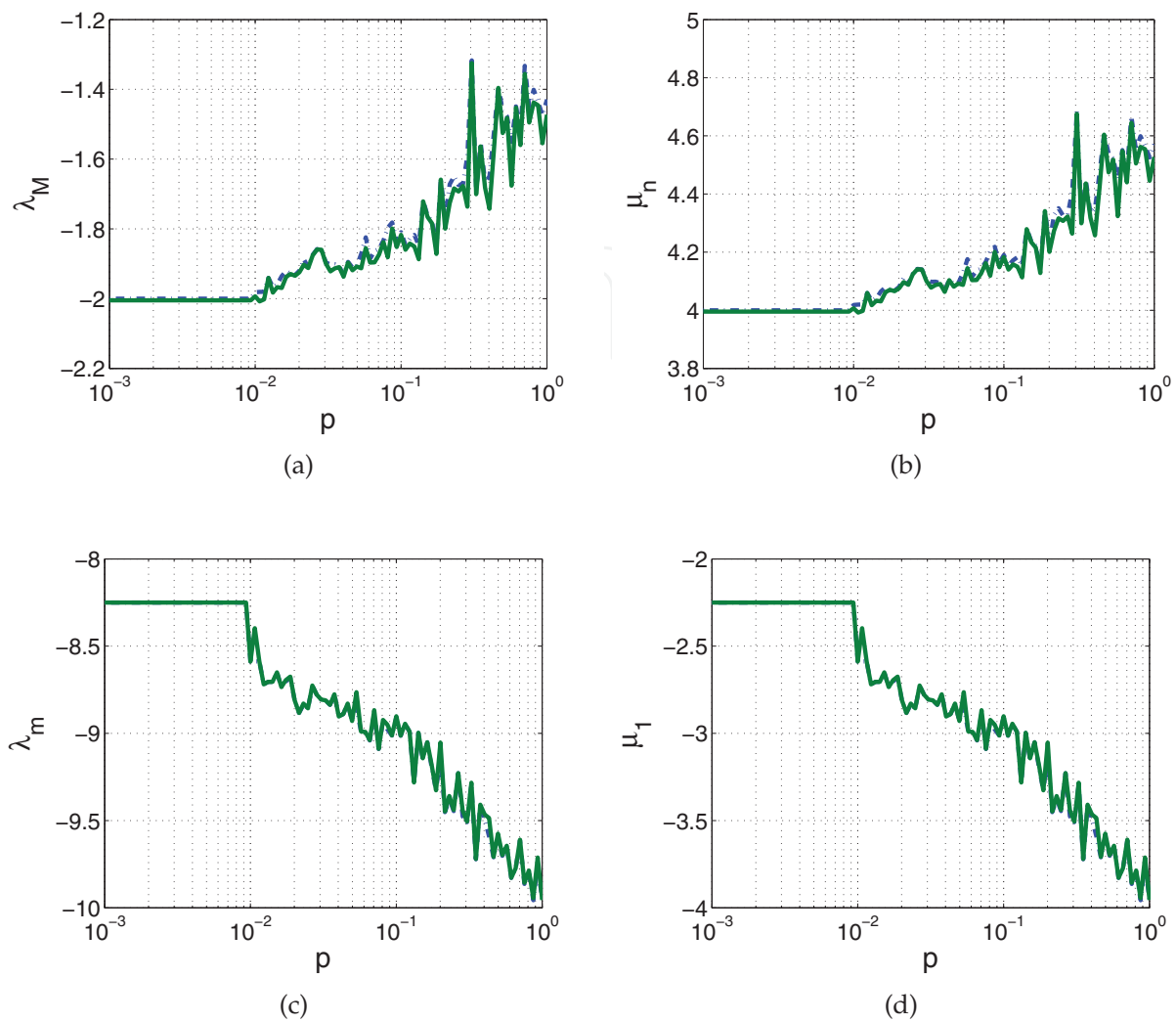


Fig. 11. Eigenvalues in the case $l = 1$. Dashed line: eigenvalue in the case of complete topology with $n = 100$; continuous line: eigenvalue in the case of node fault: (a) Maximum eigenvalue of A_g , (b) Maximum eigenvalue of U , (c) Minimum eigenvalue of A_g , (d) Minimum eigenvalue of U

the real interval with extremes the minimum and maximum A_g eigenvalues. Hence if A_g is stable, \tilde{A}_g is stable too. Moreover, the maximum eigenvalue of \tilde{A}_g is less than one of A_g . So the slowest dynamic of the system $\dot{x}(t) = \tilde{A}_g x(t)$ is faster than the system $\dot{x}(t) = A_g x(t)$.

In the follows we will show the eigenvalues of MAS dynamic in the presence of node fault. We consider MAS network with $n = 100$. We compare for each evolving network topology at each time simulation step, the maximum and minimum eigenvalues of A_g than those ones resulting with the fault of randomly chosen l nodes. Figures 11 and 12 show the eigenvalues of system dynamic for the cases $l = 1$ and $l = 50$.

Notice that as the eigenvalues of U and A_g of fault network are inside the real interval containing the eigenvalues of U and A_g of the complete graph. In Fig. 13 are shown the time evolutions of state of the complete and faulted graphs. Notice that the fault network is faster than the initial network as stated by the analysis of the spectra of A_g and \tilde{A}_g .

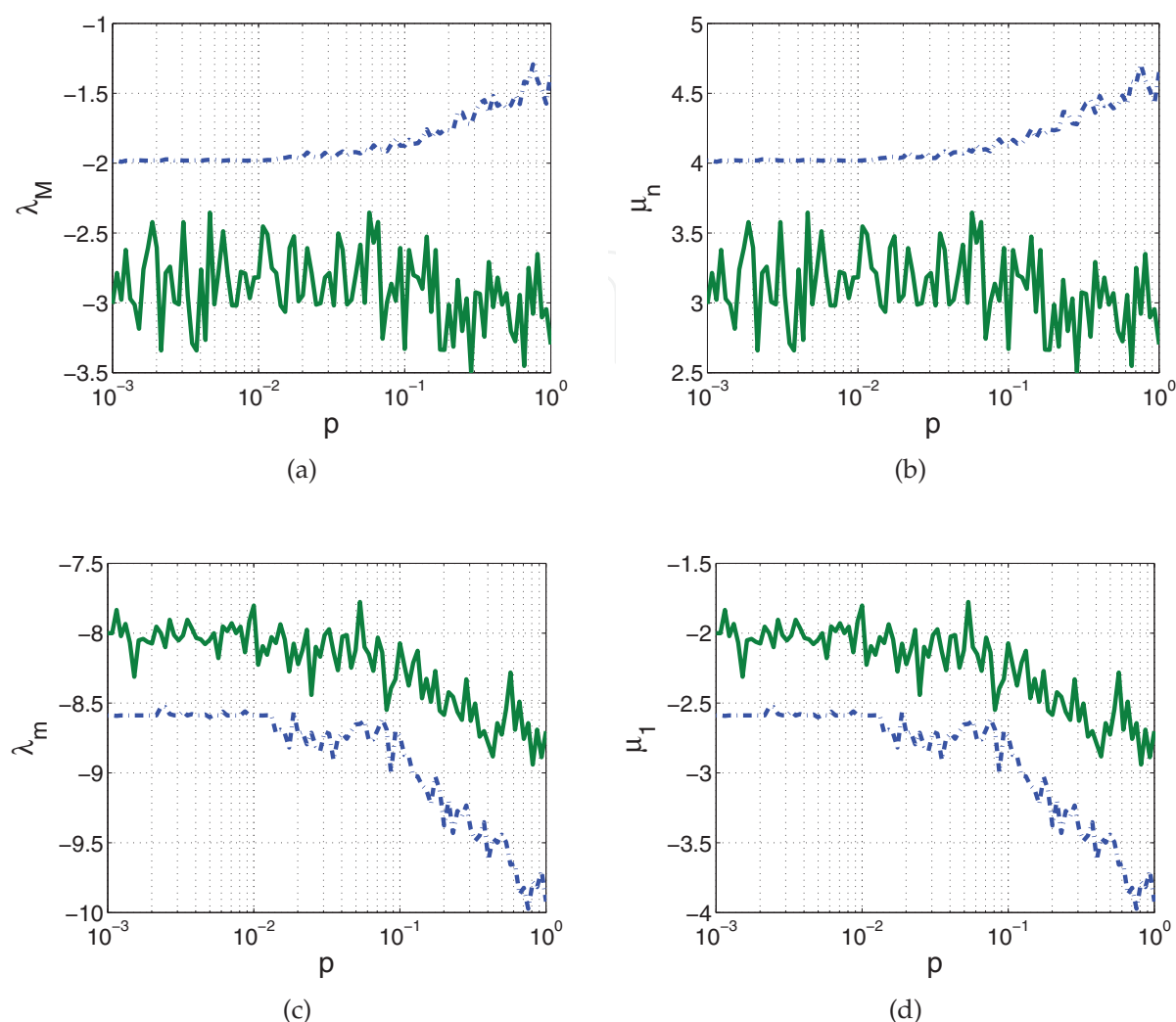


Fig. 12. Eigenvalues in the case of $l = 50$. Dashed line: eigenvalue in the case of complete topology with $n = 100$; continuous line: eigenvalue in the case of node fault: (a) Maximum eigenvalue of A_g , (b) Maximum eigenvalue of U , (c) Minimum eigenvalue of A_g , (d) Minimum eigenvalue of U

5. Conclusions

In this book chapter we have investigated the consensusability of the MASs under both the dynamic agent structure and communication topology variations. Specifically, it has given consensusability conditions of linear MASs as function of the agent dynamic structure, communication topology and coupling strength parameters. The theoretical results are given by transferring the consensusability problem to the stability analysis of LTI-MASs. Moreover, it is shown that the interplay among consensusability, node dynamic and topology must be taken into account for MASs stabilization: consensusability of MASs is assessed for all topologies, dynamic and coupling strength satisfying a pre-specified bound. From the practical point of view the consensusability conditions can be used for both the analysis and planning of MASs protocols to guarantee robust stability for a wide range of possible interconnection topologies, coupling strength and node dynamics. Also, the consensusability

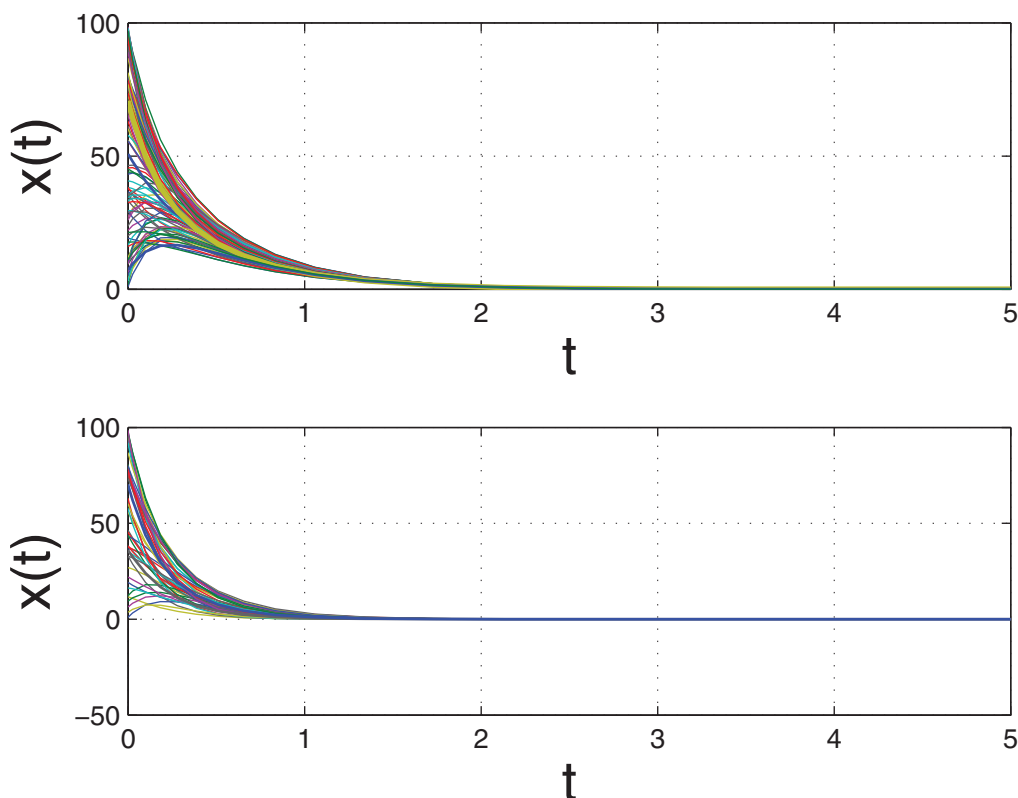


Fig. 13. Time evolution of the state variables for $l=50$: top Figure: complete graph. Bottom Figure: graph with fault.

of MAS in the presence of node faults has been analyzed. Simulation scenarios are given to validate the theoretical results.

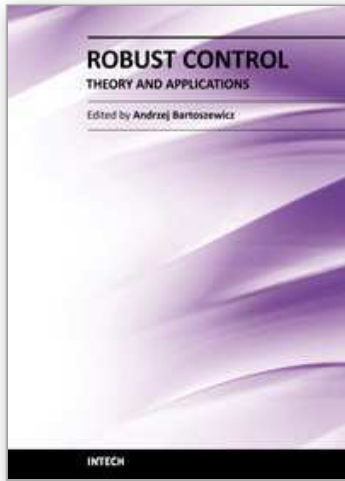
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