

*A Connective Stability  
Analysis of Complex System  
Simulation and Control via  
Multiagent Systems*

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- We present an analysis of cooperative negotiation in the area of complex complex system simulation and control

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- We present an analysis of cooperative negotiation in the area of complex complex system simulation and control
- We introduce a formalism to describe cooperative negotiation performed via multiagent systems
- We focus on a particular form of stability, called *connective stability*, for cooperative negotiation
- We present some results of a theoretical study of the connective stability of cooperative negotiation

# What is a Complex System?

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- The definition of ‘complex system’ is a debated question in literature [Astrom *et al.*, 2001]
  - Controversial definitions exist
  - A common assumption: *the behavior of a complex system depends on the interaction of several heterogeneous subsystems*
- Fields concerned with complex systems [Meyer, 1997]:
  - Physics
  - Engineering
  - Economy
  - Medicine
  - etc.

# Complex Systems in Engineering

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  - Airplane trajectory determination [Inalhan *et al.*, 2002]
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  - Robot formation management
  - etc.

# Complex Systems in Engineering

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  - Dynamic systems
  - Mathematical models
- Examples:
  - Airplane trajectory determination [Inalhan *et al.*, 2002]
  - Resource allocation [Heiskanen, 1999]
  - Robot formation management
  - etc.
- How to face complexity?
  - Adoption of different models according to several dimensions [Astrom *et altri*, 2001]:
    - Scale
    - Operating context
    - etc.
  - How can such models be harmonized?

# Classical Approaches

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  - Many concurrent models for a single phenomenon
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## Advantages

- Coexistence of many models
- Refinement of models

## Drawbacks

- There not exists integration among models
- They do not easily address highly dynamic complex systems and heterogeneous and spatial distributed models

# An Emerging Approach (1)

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- Decentralized optimization [Antsaklis, 1999]:
  - Decentralized multi-objective problem
  - $n$  independent decision makers, each one embedding a model
  - The  $n$  independent decision makers are designed to cooperate to achieve a common goal

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  - $n$  independent decision makers, each one embedding a model
  - The  $n$  independent decision makers are designed to cooperate to achieve a common goal
- Formally:
  - $\mathbf{x}_0$ : current state of the system
  - $\mathbf{u} = [u_1, \dots, u_m]$ : vector of variables
  - $f_i(\mathbf{u}) \leq 0$ : constraints of the decision maker  $i$

$$\begin{cases} \arg \min / \arg \max_{\mathbf{u}} J_i(u_1(t), \dots, u_n(t), \mathbf{x}_0) \quad \forall i \\ \text{s.t. } f_i(\mathbf{u}) \leq 0 \end{cases}$$

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## Open Issues

- How to face dynamicity of complex system?
  - Examples: new airplanes in an airspace, new resource users, new robots in a formation, etc.
- How to face heterogeneity and spatial distribution?
  - Examples: airplanes, remote resource users, robots, etc.

# A Solution for Decentralized Optim.

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- Cooperative negotiation via multiagent systems [Palm, 2004]
  - *Multiagent system* is a paradigm to model dynamic, heterogeneous, and spatial distributed entities [Weiss, 1999]
  - *Cooperative negotiation* is an optimization technique for reconfigurable and scalable systems [Bingam and Du, 2003]

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  - *Cooperative negotiation* is an optimization technique for reconfigurable and scalable systems [Bingam and Du, 2003]
- Peculiarities:
  - High formal degree (due to mathematical models)
  - High decentralization degree (due to reconfigurability and scalability)

# Cooperative Negotiation

- Classic decentralized optimization:

$$\begin{cases} \arg \min / \arg \max_{\mathbf{u}} J_i (u_1(t), \dots, u_n(t), \mathbf{x}_0) \quad \forall i \\ \text{s.t. } f_i(\mathbf{u}) \leq 0 \end{cases}$$

- Cooperative negotiation [Voos, Litz, 2000]:
  - Each agent embeds a  $J_i$  (utility  $\mathcal{U}_i$  in an economic approach)
  - Each agent performs an individual optimization (*agent optimization*)
  - All the agents, via a mediator, negotiate to reach a common agreement (*agency optimization*)

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- $\mathbf{p}_{e \rightarrow i}^t$  the counter-proposal of mediator  $e$  to agent  $i$  at time  $t$
- A negotiation session is a sequence of interleaved proposals of the agents to the mediator and counter-proposals of mediator to agents:

$$\mathbf{p}_{i \rightarrow e}^0 \succ \mathbf{p}_{e \rightarrow i}^0 \succ \mathbf{p}_{i \rightarrow e}^1 \succ \cdots \succ \mathbf{p}_{e \rightarrow i}^\tau$$

# Desirable Properties

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- Stability (does the negotiation always reach an agreement?)
  - Classical game theory cannot be adopted to design stable negotiations, but only to analyze them in particular simple cases [Roth, 1985]
  - Classical game theory cannot be adopted in presence of high decentralization

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- Stability (does the negotiation always reach an agreement?)
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  - Classical game theory cannot be adopted in presence of high decentralization
- Optimality (is the agreement optimal with respect to the agency desires?)
- Real-Time (does the agreement converge within a temporal dead-line?)
- Adaptation (can the negotiation mechanism adapt to the environmental modifications?)

# Definition of Stability

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- A succession of offers  $\mathbf{p}_{i \rightarrow e}^t$  converge to  $\bar{\mathbf{p}}_i \in A_i$  when  $\exists \bar{t} > 0$  such that  $\mathbf{p}_{i \rightarrow e}^t = \bar{\mathbf{p}}_i$  for  $t > \bar{t}$

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- A negotiation session is *strongly stable* when the succession of offers  $\mathbf{p}_{i \rightarrow e}^t$  converge to  $\bar{\mathbf{p}} \in A_i$ , for all agents  $i$ 
  - If a negotiation session is strongly stable, then the agents involved are guaranteed to converge to an agreement

# Elements Affecting Stability

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- The stability of a negotiation session depends on:
  - Utility functions  $\{\mathcal{U}_i\}$
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  - Agreement function  $\mathcal{A}$

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- The stability of a negotiation session depends on:
  - Utility functions  $\{U_i\}$
  - Negotiation functions  $\{F_i\}$
  - Agreement function  $\mathcal{A}$
- Additional problem: agents can continuously connect to and disconnect from the system
  - The stability must be studied at *runtime* whenever a *modification of the network of agents* occurs
  - In complex systems, dynamic systems, and systems that require a low latency the stability usually cannot be studied at runtime

# Introduction to Connective Stability

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- *Connective stability* is studied for dynamic systems subject to structural perturbations [Siljak, 1972]

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# Introduction to Connective Stability

- *Connective stability* is studied for dynamic systems subject to structural perturbations [Siljak, 1972]
- Multiagent systems can be considered as particular dynamic systems subject to structural perturbations (agents can connect and disconnect)
- *A cooperative negotiation is connective stable if it is stable independently from the connected agents*
  - More formally: a negotiation session is *connective strongly stable* when the succession of proposals  $p_{i \rightarrow e}^t$  converge to  $\bar{p} \in A_i$ , for all agents  $i$  independently from the connected agents

# Connective Stability Study

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- Our aim:
  - Finding general constraints that assure the connective stability of the negotiation for both analysis and design

# Connective Stability Study

- Our aim:
  - Finding general constraints that assure the connective stability of the negotiation for both analysis and design
- Our proposed methodology:
  1. Cast a negotiation session into an equivalent dynamic system with discrete time  $t$ , state  $\mathbf{P}^t = [\mathbf{p}_{1 \rightarrow e}^t \ \mathbf{p}_{2 \rightarrow e}^t \ \dots]$ , and transfer function:

$$\mathbf{p}_{i \rightarrow e}^{t+1} = \mathcal{F}_i(\dots)$$

2. Use the Lyapunov criterion [Lasalle and Lefschetz, 1961]

# Lyapunov Criterion

- Lyapunov criterion: *if there exists a smooth positive (equal to 0 in the equilibrium) monotonically decreasing function  $\mathcal{L}$  – called Lyapunov function – of the state  $\mathbf{x}$  of a dynamic system, then the system is asymptotically stable*
  1.  $\mathcal{L}(\mathbf{x}) = 0$  if and only if  $\mathbf{x}$  is the equilibrium of the dynamic system
  2.  $\mathcal{L}(\mathbf{x}) > 0$  for all other  $\mathbf{x}$
  3.  $\mathcal{L}(\mathbf{x}^t) > \mathcal{L}(\mathbf{x}^{t+1})$  for all  $\mathbf{x}$

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  3.  $\mathcal{L}(\mathbf{x}^t) > \mathcal{L}(\mathbf{x}^{t+1})$  for all  $\mathbf{x}$
- To study stability with the Lyapunov criterion, we have to find a function that satisfies the Lyapunov hypotheses for a given dynamic system
  1. We propose Lyapunov candidate functions  $\mathcal{L}$
  2. We determine the constraints for the negotiation that allow  $\mathcal{L}$  to satisfy the Lyapunov hypotheses

# Connective Stability Criterion

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- Given a generic  $\mathbf{o}_i^t \in A_i$ , the scalar  $\Gamma_i^t = \frac{\|\mathbf{p}_{i \rightarrow e}^{t+1} - \mathbf{o}_i^t\|}{\|\mathbf{p}_{i \rightarrow e}^t - \mathbf{o}_i^t\|}$  is called *contraction factor* for the agent  $i$  at the time  $t$

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- **Theorem:** Called  $n$  the number of agents, if  $\forall i \Gamma_i^t < \sqrt{\frac{n-1}{n}}$  for all  $t > \bar{t}$ , then the negotiation is connective strongly stable

# Proof (1)

- Lyapunov candidate function:

$$\mathcal{L}(\mathbf{P}^t) = \sum_{i=1}^n \sum_{j=1}^n \|\mathbf{p}_i^t - \mathbf{p}_j^t\|^2$$

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  1. It is equal to zero only in the case where  $\forall i, j : \mathbf{p}_i^t = \mathbf{p}_j^t$  (the equilibrium of the dynamic system)
  2. It is elsewhere strictly positive since it is a sum of positive terms
  3. In order to prove that it is monotonically decreasing, we must determine the constraints under which:

$$\mathcal{L}(\mathbf{P}^{t+1}) < \mathcal{L}(\mathbf{P}^t)$$

# Proof (2)

- Introducing a vector  $\mathbf{o}^t$ :

$$\begin{aligned}\mathcal{L}(\mathbf{P}^t) &= \sum_{i=1}^n \sum_{j=1}^n \|\mathbf{p}_i^t - \mathbf{p}_j^t\|^2 = \sum_{i=1}^n \sum_{j=1}^n \|(\mathbf{p}_i^t - \mathbf{o}^t) - (\mathbf{p}_j^t - \mathbf{o}^t)\|^2 = \dots \\ &\dots = 2n \sum_{i=1}^n \|\mathbf{p}_i^t - \mathbf{o}^t\|^2 - 2 \left\| \sum_{i=1}^n (\mathbf{p}_i^t - \mathbf{o}^t) \right\|^2\end{aligned}$$

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- Similarly:

$$\mathcal{L}(\mathbf{P}^{t+1}) = 2n \sum_{i=1}^n \|\mathbf{p}_i^{t+1} - \mathbf{o}^t\|^2 - 2 \left\| \sum_{i=1}^n (\mathbf{p}_i^{t+1} - \mathbf{o}^t) \right\|^2$$

# Proof (3)

- After some math, we obtain:

$$\mathcal{L}(\mathbf{P}^{t+1}) - \mathcal{L}(\mathbf{P}^t) \leq 2 \sum_{i=1}^n \|\mathbf{p}_i^t - \mathbf{o}^t\|^2 (n ((\Gamma_i^t)^2 - 1) + 1)$$

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- To have  $\mathcal{L}(\mathbf{P}^{t+1}) - \mathcal{L}(\mathbf{P}^t) < 0$  ( $\mathcal{L}$  monotonically decreasing):

$$\forall i, t: n((\Gamma_i^t)^2 - 1) + 1 < 0 \rightarrow \max_{t > \bar{t}} \Gamma^t < \sqrt{\frac{n-1}{n}} \quad \square$$

	<b>n</b>						
	2	3	4	5	10	20	100
$\sqrt{\frac{n-1}{n}}$	0.707	0.816	0.866	0.894	0.948	0.974	0.995

VALUES OF THE CONSTRAINT VALUE IN FUNCTION OF THE NUMBER  $n$  OF AGENTS

# A Particular Case

- The agreement  $\mathbf{a}^t$  is the weighted average  $\mathbf{m}^t$

$$\mathbf{a}^t = \mathbf{m}^t = \frac{\sum_{i=1}^n \mathbf{p}_{i \rightarrow e}^t \cdot \omega_i(\mathbf{p}_{i \rightarrow e}^t)}{\sum_{i=1}^n \omega_i(\mathbf{p}_{i \rightarrow e}^t)}$$

- A negotiation among  $n$  agents is equivalent to a negotiation among  $n + 1$  agents where the agent  $n + 1$  has  $\omega_{n+1} = 0$  during the negotiation
- **Corollary:** A negotiation among any number of agents with agreement  $\mathbf{a}^t = \mathbf{m}^t$  is connective strongly stable if  $\Gamma^t < 1$  for all  $t > \bar{t}$

# Conclusions

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- Cooperative negotiation via multiagent systems can effectively address reconfigurable and scalable decentralized optimization problems
- Connective stability theory is significant for cooperative negotiation via multiagent systems
- The proposed methodology can be generalized to other negotiation protocols
- Future works:
  - Apply the criterion to other negotiation protocols
  - Study the real-time properties of our negotiation protocol