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# Set Coordination of multi-agent systems

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## Outline



- Introduction
- Preliminaries
- Containment
- Set optimization
- Conclusions



#### 1. Multi-agent Systems

- Agent: subsystem with sensors and actuators
- In Nature: particles, neural network, ecological system, metabolic system...
- In Society: urban development, opinion dynamics, transportation network...
- In Engineering: power grid, wired/wireless communication, sensor network, robotic network, software system...



#### Interaction → Graph



Graph for the interaction topology of agents  $\rightarrow$ Laplacian or stochastic matrices

Directed graph: directed link (without self-loop)
Undirected graph →
bidirectional graph (undirected connections but maybe different weights)



Link (information flow): directed or undirected

#### **Consensus: a simple case**



Agent dynamics:  $dx_i/dt = u_i$  i = 1,...2Leader (or desired position):  $x_0$ 

Neighbor-based communication  $(N_i:$  the neighbor set of agent *i*)

Neighbor Graph

Distributed control:  $u_i = \sum_j (x_j - x_i), j \in N_i$ 

Multi-agent consensus (agreement, synchronization): •Leader-following:  $x_i - x_0 \rightarrow 0$ •Leaderless:  $x_i - x_j \rightarrow 0$ 



# **Multi-agent Coordination**

- Coordination Problems:
  - Consensus (agreement, synchronization): a basic problem
  - Rendezvous, flocking, swarm
  - Formation, sweeping …
- Coordination  $\rightarrow$  connectivity
  - Interconnection topology: information flow
  - Switching topologies: joint connection
  - Leader-following or leaderless





## **Distributed analysis & control**



Hierarchy and Interaction

## **Set coordination**

- What happens if a set Ω gets involved with the coordination of a group of agents?
- How to give distributed algorithms for set tracking or set optimization?
- What if there are multiple leaders? ...



 $v_2$ 

#### Motivation



Migration or swarm to food source or nest site



Couzin, et al., Leadership by Numbers Nature, vol 433., Feb. 2005

**Results:** The effectiveness of the leaders (informed agents)

#### **Motivation 2**



Agents (sheep) are contained in a region spanned by a group of leaders (sheepdogs) → containment control:

- Multiple leaders: form a moving polytope (convex target set)
- Multiple followers (agents): aim at reaching the convex leader-set.



#### **Motivation 3**



- Practical target set (due to inaccuracy in measurement, uncertainty in environment...) for coordination or decision making
- Distributed set optimization, optimal consensus within a convex set



#### Set coordination problems



- Flocking to a given position → Drive a group of agents into a target convex set Ω
- Leader-following → multi-leader following → containment control: contain the agents in a convex set spanned by leaders
- Multi-agent consensus → set optimization: consensus together with convex optimization

#### Difficulties



- Centralized design → Distributed design: neighbor-based rule, not completely connected
- Conventional stability → set stability (nonsmoothness)
- Switching interaction topology (nonsmoothness): common Lyapunov function



# **Distributed algorithm with set**



#### Convex set: Convex analysis

#### **Related results**



- Stationary set: Couzin, et al. Nature, 2005; Lin, et al, IEEE TAC'05 (a segment); Shi, Hong, Automatica'09 (any convex set); ...
- Moving-leader containment: Ji et al, IEEE TAC'07; Ren et al, IEEE TAC'10; Shi, Hong, et al IEEE TAC'12; Lou, Hong, Automatica'12
- Convex optimization with distributed consensus: Johansson et al, IEEE CDC'08; Nedic et al, IEEE TAC'09, 10; Shi, Johansson, Hong, (IEEE TAC conditionally accepted 2012), Lou et al, WCICA'12.

#### 2. Preliminaries



Graph theory: the interaction topology of agents (which is important especially in the homogenous agent case) → algebraic graph theory: Laplacian or stochastic matrices.

- Directed graph (without self-loop)
- Undirected graph → bidirectional graph (undirected connections but weights may be different)

# Connectivity



- Joint connection: union graph in  $[t, \infty)$  is connected for any *t*: a necessary condition
- Uniform joint connection:  $\exists T$ , union graph on [t,t+T] is connected for any *t*.



#### **Convex set**



• *K* is a convex set if, for  $0 < \lambda < 1$ 

$$(1-\lambda)x + \lambda y \in K \qquad x \in K, y \in K$$

• d(x,K): distance between set K and x

$$||x||_K \triangleq \inf\{||x-y|||y \in K\}$$

## **Dini derivative**



# • Dini derivative: for a continuous function h $D^+h(t) = \limsup_{s \to 0^+} \frac{h(t+s) - h(t)}{s}$

• Dini derivative for a switching function V: if  $V(t,x) = \max_{i=1,2,\dots,n} V_i(t,x)$   $D^+V(t,x(t)) = \max_{i\in\mathcal{I}(t)} \dot{V}_i(t,x(t))$  $\mathcal{I}(t) = \{i \in \{1,2,\dots,n\} : V(t,x(t)) = V_i(t,x(t))\}$ 

# Input-to-state Stability (ISS)

- ISS proposed by Sontag for nonlinear control systems, widely used in stabilization & robust control (for interconnected system)
- dx/dt = f(x, u), x: state; u: input
- The system is ISS if  $//x(t)//\leq \beta(//x(0))//, t) + \gamma(//u//_{\infty})$ , with *K*-function  $\gamma$  and *KL*-function  $\beta$
- ISS  $\rightarrow$  bounded input bounded output; ISS  $\rightarrow$  asymptotic stability if no input;

#### **Related discussion**



- Integral ISS (iISS): the system is integral ISS if  $||x(t)|| \leq \beta(||x(0)||, t) + \int^t \gamma(||u(s)||) ds$
- Set ISS (SISS) by Sontag: static/fixed set and system without switching
- There are few ISS results for multi-agent systems with fixed topologies (Scardovi et al, IEEE CDC'09; Tanner et al, IEEE TRA'04)

Not the set case we studied for switched multi-agent systems with moving sets: new SISS & SiISS

#### 3. Containment



- Multiple leaders: form a moving polytope (convex target set)
- Multiple followers (agents): aim at reaching the convex target set.
- Tracking a moving set in a distributed way
- Bounded tracking error if the uncertainties are bounded

#### **Distributed design**



- Leaders: dy<sub>i</sub>/dt=v<sub>i</sub>(y,t), i=1,...,k; v<sub>i</sub> is the velocities
- Agents: dx<sub>i</sub>/dt=u<sub>i</sub>, i=1,...,n; u<sub>i</sub> is a neighborbased rule in a nonlinear form with uncertainty w<sub>i</sub>:

 $\sum_{\substack{j \in N_i(\sigma(t)) \\ \cdot (y_j - x_i) + w_i(t), \quad i = 1, \cdots, n}} a_{ij}(x, y, t) (x_j - x_i) + \sum_{\substack{j \in L_i(\sigma(t)) \\ \cdot (y_j - x_i) + w_i(t), \quad i = 1, \cdots, n}} b_{ij}(x, y, t)$ 

 Uncertainties: leaders with unknown velocities v; agents with disturbance w

#### **Basic setup**



- The position is measurable, but the velocity is unmeasurable
- Position information of a leader is known by an agent if and only if the agent is connected to the leader: local controller only contains position information
- The interconnection topology is timevarying (widely used in multi-agent control)

# Variable topology





 $L_i$ : moving leaders  $\rightarrow$  span a convex set  $\Omega$ 

 $F_i$ : follower agents, time-varying communication Time-varying graph  $\rightarrow$  L-connected; jointly L-connected, ...

#### **L-connection**

<u>L-connected</u>: any agent is accessible to a leader; Jointly L-connected (JLC) : union graph of interval [s,t] is Lconnected; <u>Uniformly JLC</u>:  $\exists T$ such that the union graph of [t, t+T] is Lconnected





# Set tracking and SISS

- SISS → set tracking if there is no uncertainty
- SISS with the uncertainties (v,w) as input with respect to the target set  $\Omega$
- *K*-function: γ; *KL*-function: β
- SISS:  $d(x_i, \Omega) \leq \beta(d(x_i(0), \Omega), t) + \gamma(//v, w//_{\infty})$
- SiISS (set integral input-to-state stability):  $d(x_i, \Omega) \leq \beta(d(x_i(0), \Omega), t) + \int^t \gamma(||v,w||) ds$

# Analysis



- Different from ISS, no equivalence relationship with ISS-Lyapunov function
- Constructing non-smooth Lyapunov function:  $V(x) = max_i d(x_i, \Omega)$
- Estimating the convergence rate for SISS or SiISS by finding suitable K-function γ and KL-function β.
- Estimation is done for each interval in the case of joint L-connection

#### Main results: SISS



#### Key result 1: SISS of MAS = uniformly jointly L-connected (UJLC)

Remarks:

- SISS → bounded (uncertain) input, bounded tracking error
- Completely L-connected  $\rightarrow$  SISS

## Main results: SilSS



Key result 2: Uniformly jointly L-connected → SiISS of MAS

Key result 3: For bidirectional graph, SiISS =  $[t, \infty)$  jointly L-connected for any t

Remarks:

- For our system: SISS (UJLC)  $\rightarrow$  SiISS
- Other sufficient conditions: Complete or Acyclic L-connection ... → SiISS

# Set tracking



- Set tracking (without error): the agents enter the convex target set.
- Set tracking ←
- 1. SISS + (v, w) vanishing; or
- 2. SiISS +  $\int_{-\infty}^{\infty} \gamma$  bounded
- Consistent with ISS in the conventional single nonlinear system.

# 4. Distributed set optimization



- Optimization of the whole system with global cost function when each agent optimizes its own cost.
- A hot topic: convex cost functions (corresponding to convex set); local information; topology may be switching ...
- Our result: weak connectivity condition, continuous-time system with a new method, optimization with consensus.



#### Introduction

Distributed optimization over networks

$$\min_{z \in \mathbb{R}^m} F(z) = \sum_{i=1}^N f_i(z)$$



- Each node i observes its own cost function  $f_i$  only
- Information is exchanged locally over communication graphs
- All nodes target to reach the optimal solution set  $X_0 = \arg \min_{z \in \mathbb{R}^m} F(z)$
- Applications
  - Coordination control of multi-agent systems
  - Resource allocation in wireless systems
  - Opinion dynamics in social networks

#### Methods



- Gradient-based: manipulation of the primal variable through (sub)gradient with local (environment) information ...
- Neighbor-based: optimization with local communication information
- Swarm intelligence, genetic algorithm, ... (hard to be analyzed)

#### **Our Problem**

- Agent  $dx_i/dt = u_i$  only knows the information of its own convex set  $X_i$  and its neighbor  $x_j \rightarrow$  the agents achieve consensus within  $X_0$  $(= \cap X_i$ , which is not empty)
- Distributed convex intersection (computer): Projected consensus algorithm (PCA) by Nedic et al 2010.....





#### **Optimal consensus**



Establish relationships between connectivity conditions and optimal consensus.

Basic model:

- Connectivity: uniformly jointly-connected for digraph, [t, ∞) jointly-connected for bidirectional (undirected) graph
- Agents: dx<sub>i</sub>/dt=u<sub>i</sub>, i=1,...,n; u<sub>i</sub> a nonlinear local rule = subgradient for optimization + neighbor-based rule for consensus
- The intersection set  $X_0 = \cap X_i$  is not empty
- Dwell time



## **Convex projection**





#### **Distributed Control**

Control law for node *i*:

$$\dot{x}_{i} = u_{i} = \sum_{j \in \mathcal{N}_{i}(t)} a_{ij} (x, t) (x_{j} - x_{i}) + P_{X_{i}}(x_{i}) - x_{i}$$

- A simple combination of a consensus algorithm and a projection item
- A global coordinate system is not needed
- Each node does not need to know its state

#### Main results



Result 1: Global optimal consensus of MAS ← uniformly jointly strongly connected

Result 2: In the bidirectional case, MAS achieves global optimal consensus =  $[t, \infty)$  joint connection

# **Approximate projection**

In practice, it is hard to get accurate projection (PCA)  $\rightarrow$  approximate projection with  $0 \le \theta \le \theta^* < \pi/2$  (APCA):

Optimal consensus: given APCA, there is  $x^* \in X_0$ such that

$$\lim_{k \to \infty} x_i(k) = x^*, \ i = 1, ..., n$$







 $C_{K}(v, \theta)$ -v is a cone

#### **Distributed Control**



$$\begin{array}{lll} \text{Controller:} & x_i(k+1) = \sum_{j \in \mathcal{N}_i(k)} a_{ij}(k) \left( (1 - \alpha_{j,k}) x_j(k) + \alpha_{j,k} P_j^{sa}(k) \right) \\ \text{where} & P_i^{sa}(k) \in \mathscr{P}_{X_i}^{sa}(x_i(k), \theta_k) \\ & \alpha_k^- = \min_{1 \leq i \leq n} \alpha_{i,k} \\ & \alpha_k^+ = \max_{1 \leq i \leq n} \alpha_{i,k} \end{array} \qquad \begin{array}{c} X_1 \\ & & Y_1 \\ & & & Y_1 \\ & & & & & \\ & & & & \\ & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\$$

#### Main results for APCA



Result 1: Global optimal consensus of MAS  $\leftarrow$  uniformly jointly strongly connected &

$$\sum_{k=0}^{\infty} \alpha_k^- = \infty \ \& \ \sum_{k=0}^{\infty} \alpha_k^+ \theta_k < \infty$$

Result 2:  $X_i$  is bounded and  $0 < \theta < \pi/4 \rightarrow$ 

$$\sup_{x(0)} \limsup_{k \to \infty} |x_i(k)|_{X_0} < \infty$$



#### **Numerical example**

5 4 3 Circle 1 Circle 2 2 Circle 3 x1 1 x2 - x3 0

2

1

3

5

4

6

0

-3

-2

-1

t = 0 to t = 1000

#### **5. Conclusions**



- Multi-agent system: promising
- Coordination problems beyond multi-agent consensus: set coordination, distributed optimization, ...
- Set coordination for more generalized models: higher-order systems, eventtriggered control, or stochastic or constrained models, ...



# Thank you !

