



Distributed Event-Based Control

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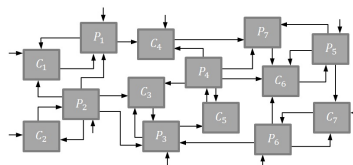
Workshop on Event-Based Control and Optimization
TUM Institute for Advanced Study, Oct 1-2, 2012

Motivation

Networked control systems have time-varying communications influencing their global performance

Network topology depends on

- Internal events: states, controls
- External events: disturbances, outages

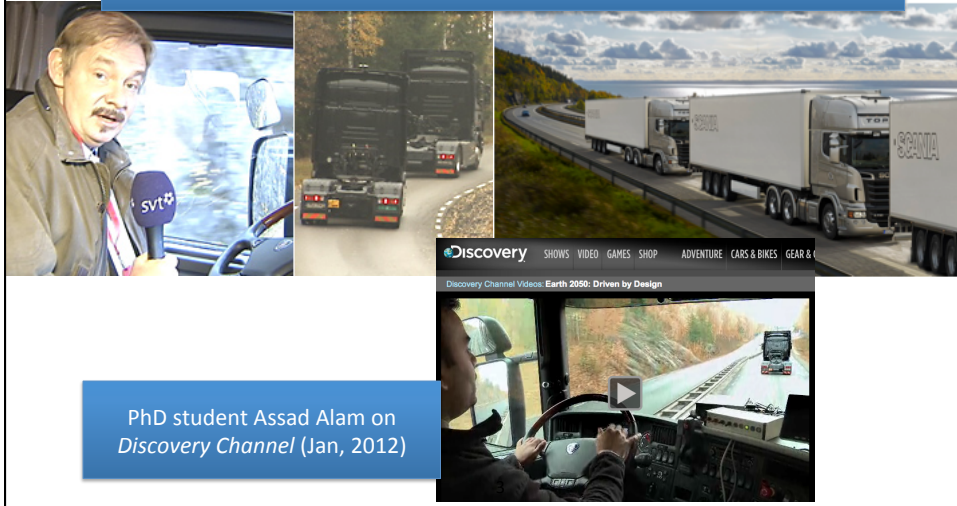


Need for integrated design of control and communication

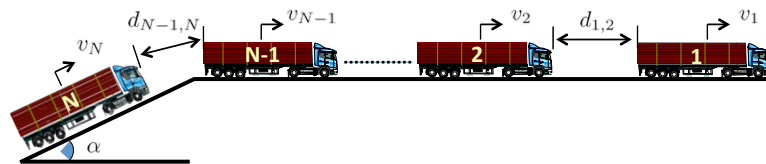


Truck Platooning for Fuel Reduction

Rapport on vehicle platooning developed by KTH and Scania (Oct, 2011)



Networked Control in Platooning



- Platooning control applications require **collaborative actions**
 - Fuel-efficient adaptive cruise controllers
 - Collaborative route planning
 - Autonomous safety maneuvers
 - Vehicles need **accurate estimates** of neighboring vehicles' states and actions
 - Control performance is tightly coupled to how well data (position, velocity, braking estimates) are communicated across the platoon
- How does the communication influence the system performance?
 - What is an efficient communication strategy for specific control tasks?



Mathematical Model

Directed graph $\mathcal{G} = (\mathcal{V}, \mathcal{E})$

Node set $\mathcal{V} = \{1, 2, \dots, n\}$

Arc $e = (i, j) \in \mathcal{E}$

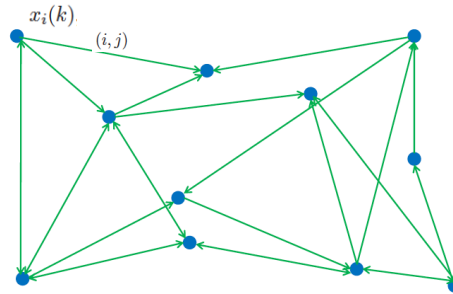
Time-varying graph process

$$\mathcal{G}_k(\omega) = (\mathcal{V}, \mathcal{E}_k(\omega)), k = 0, 1, \dots$$

To each node $i \in \mathcal{V}$, associate a scalar state $x_i(k)$

x_i updates based on own computation and neighbor information

$$\mathcal{N}_i(k) = \{j \in \mathcal{V} : (j, i) \in \mathcal{E}_k\} \cup \{i\}$$



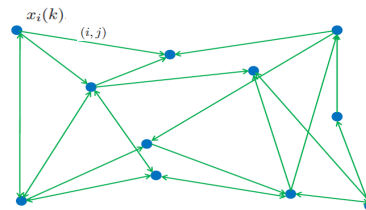
Objective

Control the states to agreement: $\lim_{k \rightarrow \infty} |x_i(k) - x_j(k)| = 0$ for all $i, j \in \mathcal{V}$

Also called *consensus*, *rendezvous*, *formation*, etc

Local update law

$$x_i(k+1) = \sum_{j \in \mathcal{N}_i(k)} a_{ij}(k) x_j(k)$$



Prototype model for a collaborative control problem with
coupled network and node dynamics

Related work on Markov chains, belief evolution, consensus algorithms, distributed control etc:

Hajnal (1958), Wolfowitz (1963), DeGroot (1974), Tsitsiklis, Bertsekas & Athans (1986), Jadbabaie, Lin & Morse (2003), Moreau (2005), Ren & Beard (2005), Golub & Jackson (2007), Cao, Anderson & Morse (2008), Acemoglu, Ozdaglar & ParandehGheib (2010), etc

A Gossip Algorithm

At each k , select a pair of nodes that “gossip”:

$$x_i(k+1) = \begin{cases} [x_i(k) + x_j(k)]/2 & \text{if } (i,j) \text{ or } (j,i) \text{ is selected} \\ x_i(k) & \text{otherwise} \end{cases}$$

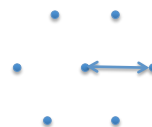
Equivalently

$$x(k+1) = A_k x(k)$$

where $A_k \in \mathcal{A}$ with

$$\mathcal{A} = \{I - (e_i - e_j)(e_i - e_j)^T/2 : i, j \in \mathcal{V}\}$$

and e_m is the unit vector



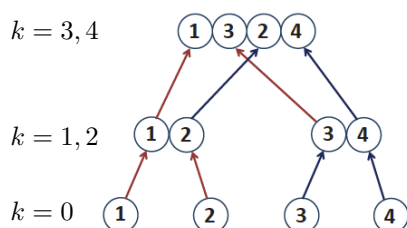
Can the gossiping pairs be selected to achieve finite-time convergence ?

Various bounds on the convergence time to asymptotic consensus, e.g., Karp et al. (2000), Kempe et al. (2003), Boyd et al., (2006), Shah (2008), Liu et al. (2011)

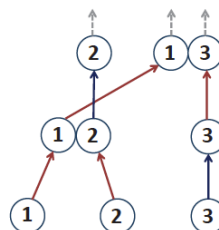
Gossiping Convergence: Examples

$$x_i(k+1) = \begin{cases} [x_i(k) + x_j(k)]/2 & \text{if } (i,j) \text{ or } (j,i) \text{ is selected} \\ x_i(k) & \text{otherwise} \end{cases}$$

Convergence in 4 steps for $n=4$ nodes

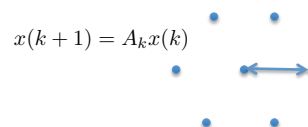


No finite-time convergence for $n=3$ nodes



Finite-Time Convergence of Gossiping

Theorem: There exists a deterministic gossip algorithm $\{A_k\}_{k=0}^{\infty}$ that ensures global finite-time convergence if and only if there exists an integer $m \geq 0$ such that the number of nodes $n = 2^m$.

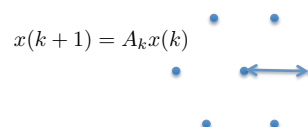


The proof is constructive and provides a gossip algorithm reaching (fastest?) global convergence in $(n \log_2 n)/2$ steps

Shi et al. (2012)

Impossibility of Finite-Time Convergence

Theorem: Suppose there exists no integer $m \geq 0$ such that $n = 2^m$. Then, for almost all initial values, it is impossible to find a gossip algorithm $\{A_k\}_{k=0}^{\infty}$ that reaches finite-time convergence.

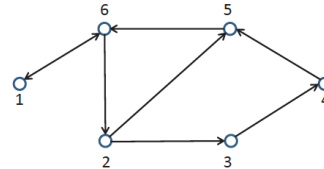


Shi et al. (2012)

Distributed Averaging and Maximizing

$$x_i(k+1) = \eta_k x_i(k) + \alpha_k \min_{j \in \mathcal{N}_i(k)} x_j(k) + (1 - \eta_k - \alpha_k) \max_{j \in \mathcal{N}_i(k)} x_j(k)$$

$$\eta_k \in [0, 1] \text{ and } \alpha_k \in [0, 1 - \eta_k]$$



$\eta_k \equiv 0, \alpha_k \equiv 0$: distributed maximizing

$\eta_k \equiv 0, \alpha_k \equiv 1$: distributed minimizing

$\eta_k \in (0, 1], \alpha_k \in [0, 1 - \eta_k]$: distributed weighted averaging

Impossibilities of Convergence

$$x_i(k+1) = \eta_k x_i(k) + \alpha_k \min_{j \in \mathcal{N}_i(k)} x_j(k) + (1 - \eta_k - \alpha_k) \max_{j \in \mathcal{N}_i(k)} x_j(k)$$

Averaging algorithms: $\eta_k \in (0, 1], \alpha_k \in [0, 1 - \eta_k]$

Theorem: For every averaging algorithm, **finite-time** convergence fails for all initial conditions except for the consensus manifold.

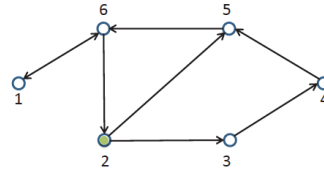
Theorem: For every averaging algorithm, **asymptotic** convergence fails for all initial conditions except for the consensus manifold if $\sum_{k=0}^{\infty} (1 - \eta_k) < \infty$.

These results are independent of the network topology

Convergence of Maximizing Algorithms

$$x_i(k+1) = \eta_k x_i(k) + \alpha_k \min_{j \in \mathcal{N}_i(k)} x_j(k) + (1 - \eta_k - \alpha_k) \max_{j \in \mathcal{N}_i(k)} x_j(k)$$

Maximizing algorithms: $\nu_k \equiv \alpha_k \equiv 0$



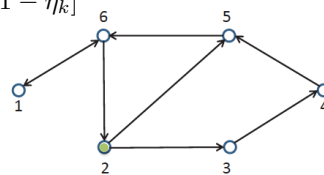
Theorem: Suppose $\mathcal{G}_k \equiv \mathcal{G}_*$ is a fixed graph. Global finite-time convergence is achieved if and only if \mathcal{G}_* is strongly connected.

Shi & J (2012)

Convergence of Averaging Algorithms

$$x_i(k+1) = \eta_k x_i(k) + \alpha_k \min_{j \in \mathcal{N}_i(k)} x_j(k) + (1 - \eta_k - \alpha_k) \max_{j \in \mathcal{N}_i(k)} x_j(k)$$

Averaging algorithms: $\eta_k \in (0, 1]$, $\alpha_k \in [0, 1 - \eta_k]$



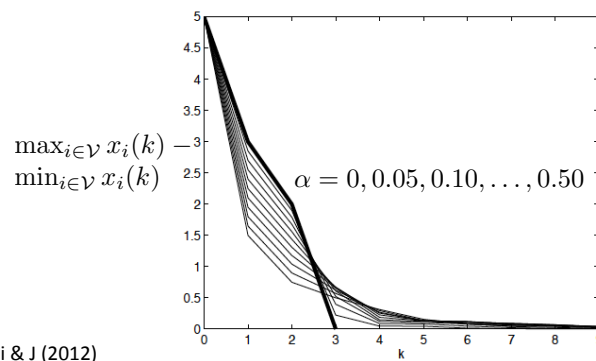
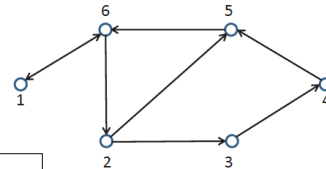
Theorem: Suppose $\mathcal{G}_k \equiv \mathcal{G}_*$ is a fixed graph and $\alpha_k \equiv \alpha > 0$. Global asymptotic convergence is achieved if and only if \mathcal{G}_* has a root.

Shi & J (2012)

Example

$$x_i(k+1) = \alpha \min_{j \in \mathcal{N}_i(k)} x_j(k) + (1 - \alpha) \max_{j \in \mathcal{N}_i(k)} x_j(k)$$

- $0 < \alpha < 1$: global **asymptotic** consensus
- $\alpha = 0$ or $\alpha = 1$: global **finite-time** consensus



Shi & J (2012)

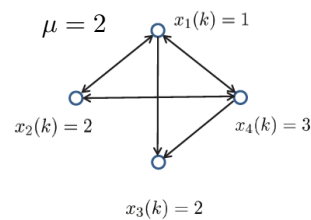
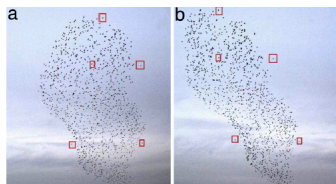
State-Dependent Nearest-Value Graphs

Fix positive integer μ

Neighbors of node $i \in \mathcal{V}$ are nodes in the union of

$$\mathcal{N}_i^-(k) = \{\text{nearest } \mu \text{ neighbors } j \in \mathcal{V} \text{ with } x_j(k) < x_i(k) \text{ and distinct values}\}$$

$$\mathcal{N}_i^+(k) = \{\text{nearest } \mu \text{ neighbors } j \in \mathcal{V} \text{ with } x_j(k) > x_i(k) \text{ and distinct values}\}$$



Motivated from recent studies of starlings collective behavior [Ballerini et al., PNAS, 2008]

Finite-time Convergence

$$x_i(k+1) = \eta_k x_i(k) + \alpha_k \min_{j \in \mathcal{N}_i(k)} x_j(k) + (1 - \eta_k - \alpha_k) \max_{j \in \mathcal{N}_i(k)} x_j(k)$$

Theorem: Consider a nearest-value graph and an averaging algorithm with $\eta_k \equiv 0$ and $\alpha_k \in (0, 1)$.

- (i) If $n \leq 2\mu$, then global finite-time consensus is achieved.
- (ii) If $n > 2\mu$, then no finite-time consensus is achieved for almost all initial conditions.

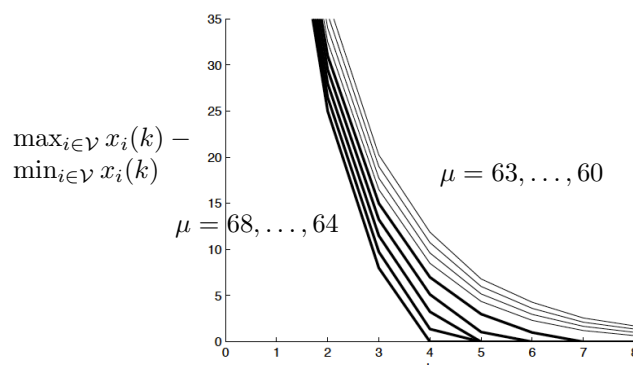
Finite-time convergence only with sufficiently many neighbors

Shi & J (2012)

Example

$$x_i(k+1) = \alpha \min_{j \in \mathcal{N}_i(k)} x_j(k) + (1 - \alpha) \max_{j \in \mathcal{N}_i(k)} x_j(k)$$

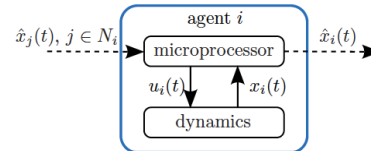
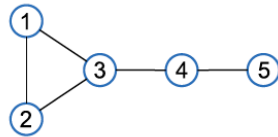
$n = 128$ nodes and $\alpha = 1/2$



Finite-time convergence for only for many neighbors: $\mu \geq 64$

Shi & J (2012)

State-dependent Event-based Scheduling of Measurement Communications



$$\dot{x}_i(t) = u_i(t)$$

$$u_i(t) = - \sum_{j \in N_i} (\hat{x}_i(t) - \hat{x}_j(t))$$

Event-based broadcasting

$$\hat{x}_i(t) = x_i(t_k^i), t \in [t_k^i, t_{k+1}^i[$$

$$0 \leq t_0^i \leq t_1^i \leq t_2^i \leq \dots$$

$$t_{k+1}^i = \inf\{t : t > t_k^i, f_i(t) > 0\}$$

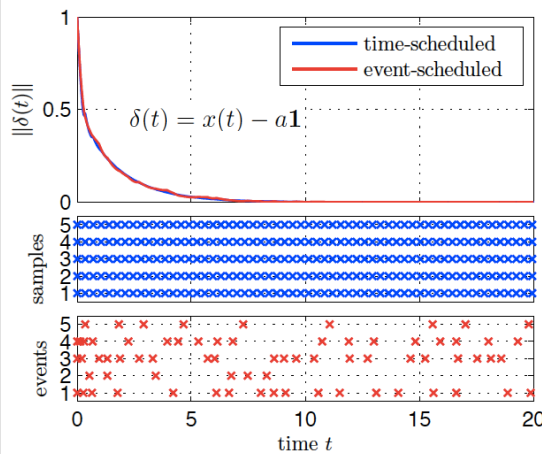
$$f_i(t, e_i(t)) = |e_i(t)| - (c_0 + c_1 e^{-\alpha t})$$

$$e_i(t) = \hat{x}_i(t) - x_i(t)$$

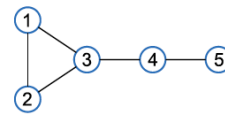
Practical consensus is achieved if $0 < \alpha < \lambda_2(L)$

Seyboth et al. (2011)

Event-based vs Periodic Communication



Graph:



Sampling periods:

■ Time-scheduling:

$$\tau_s = 0.350$$

$$\tau_{max} = 0.480$$

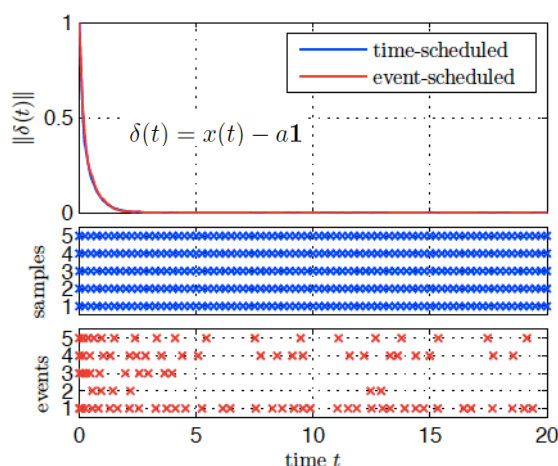
■ Event-scheduling:

$$\tau_{mean} = 1.389$$

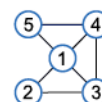
τ_{max} : largest stabilizing sampling period, see [G. Xie et al., ACC2009](#)

Seyboth et al. (2011)

Event-based vs Periodic Communication



Graph:



Sampling periods:

■ Time-scheduling:
 $\tau_s = 0.250$
 $\tau_{max} = 0.400$

■ Event-scheduling:
 $\tau_{mean} = 1.053$

τ_{max} : largest stabilizing sampling period, see *G. Xie et al., ACC2009*

Seyboth et al. (2011)

Conclusions

- Controls and communications are coupled in many applications
- Fundamental limits for some control objectives and network protocols
 - Finite-time and asymptotic convergence, gossiping and averaging
- Tradeoffs between communication capacity and control computations

Extensions

- Stochastic dynamics and networks
- PHY, MAC, NET models



<http://www.ee.kth.se/~kallej>

