



Event-Based Control of Multi-Agent Systems

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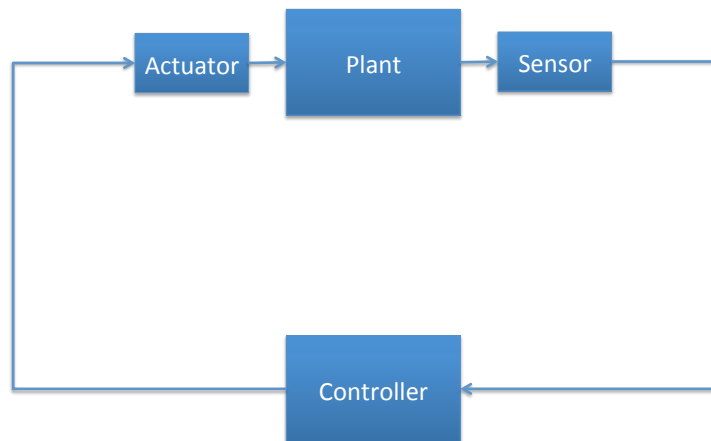
Dimos Dimarogonas, Henrik Sandberg

together with collaborations and inspiring discussions with more.

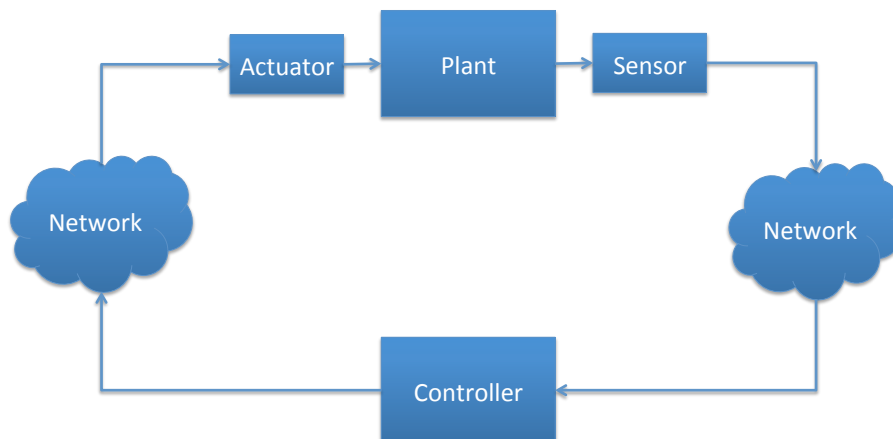
Funding sources:



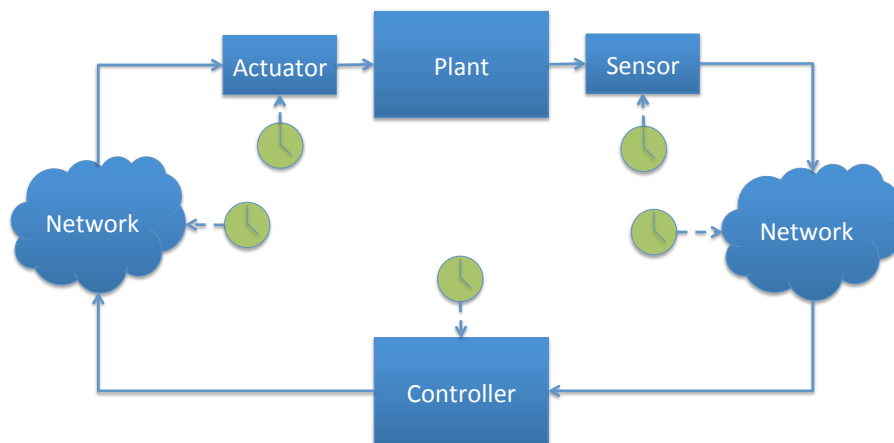
Feedback Control System



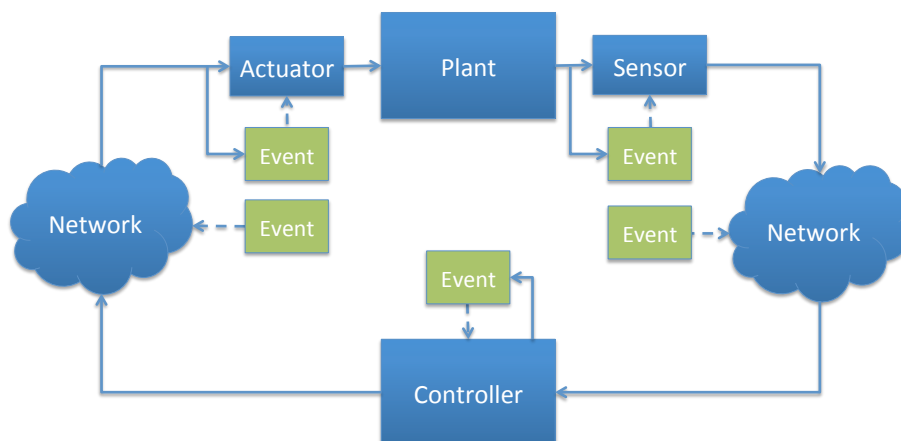
Networked Control System



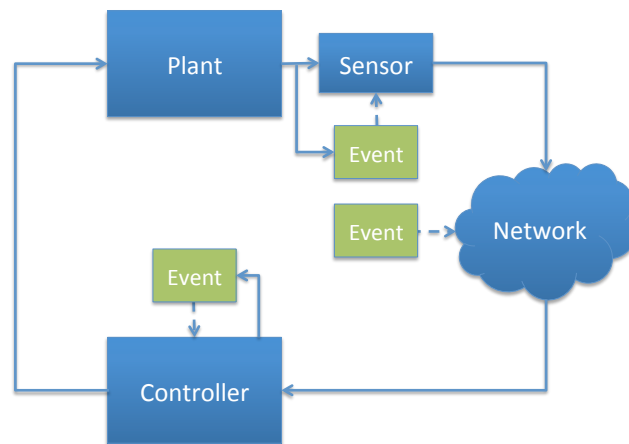
Time-Triggered Control System



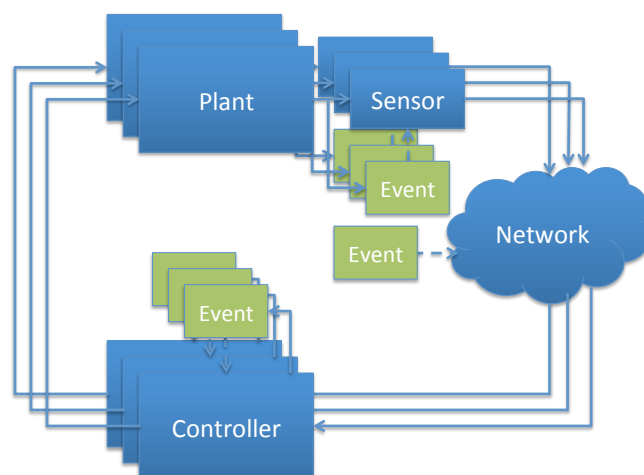
Event-Based Control System



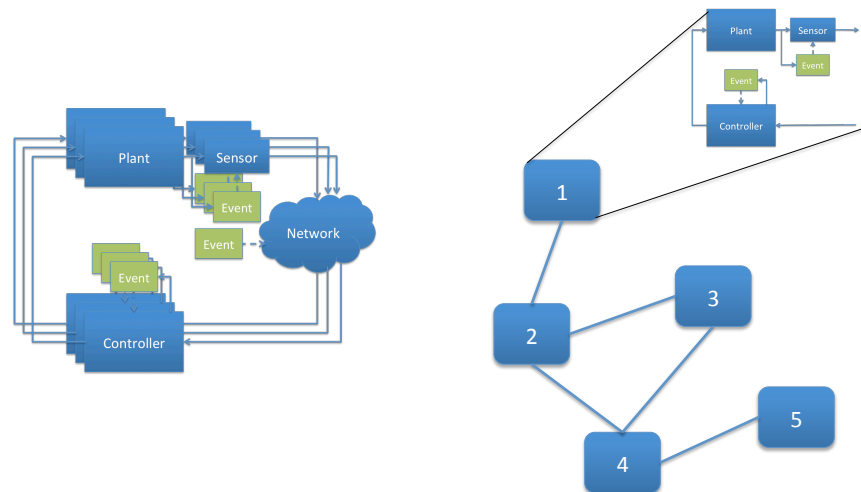
Event-Based Control System



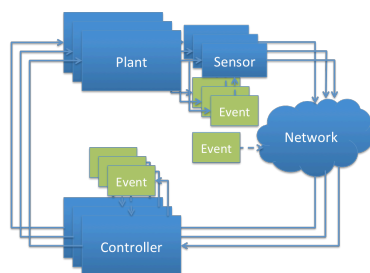
Event-Based Control System



Event-Based Multi-Agent System



Goal: Guarantee Control Performance under Limited Resources



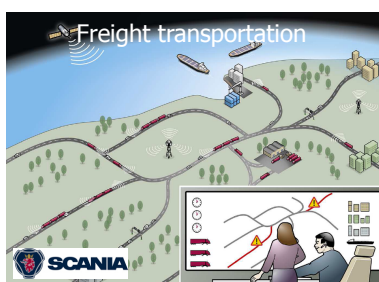
Resources

- Sensing
- Sensor communication
- Network
- Actuation
- (Computing)

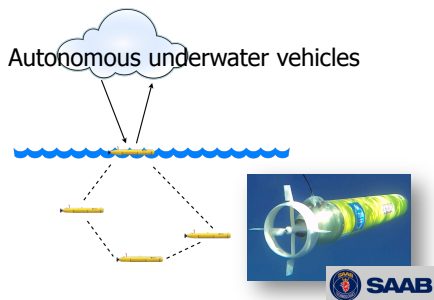
Outline

- Introduction
- Motivating applications
- Optimal event-based control
- Distributed event-based control
- Implementation aspects
- Conclusions

Motivating Applications

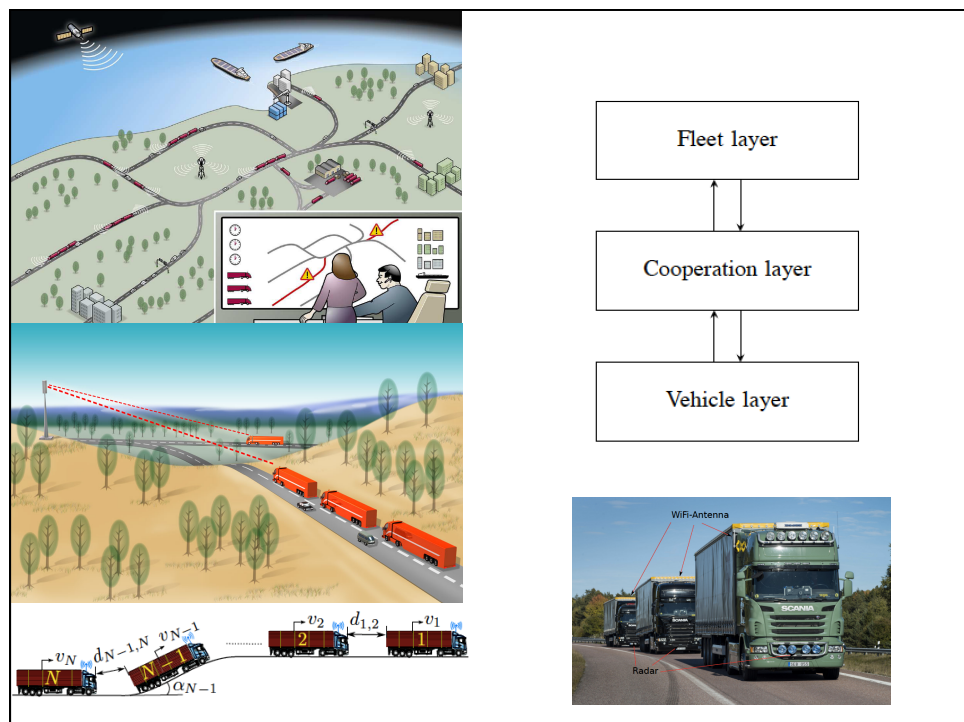
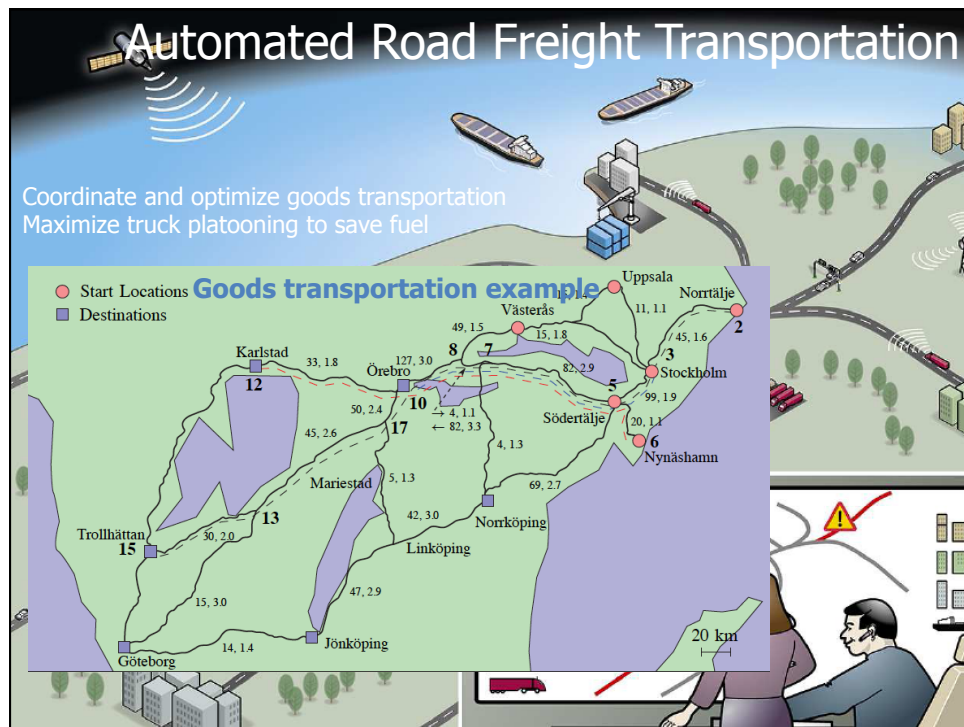


Autonomous underwater vehicles

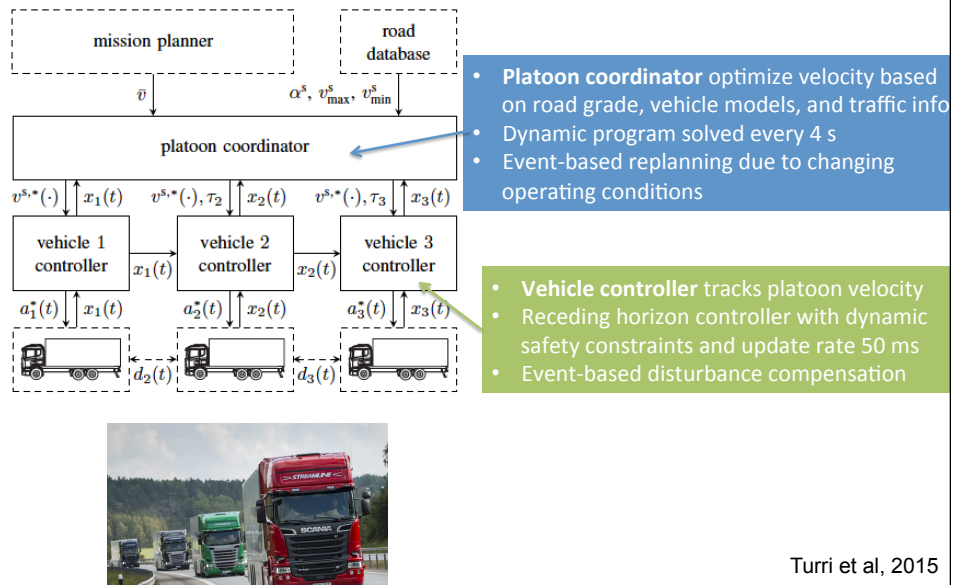


Wireless industrial automation





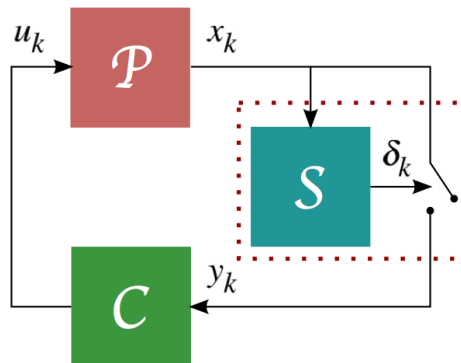
Event-based Control of Vehicle Platoon



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Optimal Event-Generation and Control



Stochastic Control Formulation

Plant:

$$x_{k+1} = Ax_k + Bu_k + w_k$$

Scheduler:

$$\delta_k = f_k(\mathbb{I}_k^S) \in \{0, 1\}$$

$$\mathbb{I}_k^S = [\{x\}_0^k, \{y\}_0^{k-1}, \{\delta\}_0^{k-1}, \{u\}_0^{k-1}]$$

Controller:

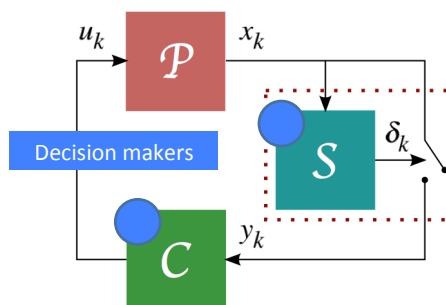
$$u_k = g_k(\mathbb{I}_k^C)$$

$$\mathbb{I}_k^C = [\{y\}_0^k, \{\delta\}_0^k, \{u\}_0^{k-1}]$$

Cost criterion:

$$J(f, g) = \mathbb{E}[x_N^T Q_0 x_N + \sum_{s=0}^{N-1} (x_s^T Q_1 x_s + u_s^T Q_2 u_s)]$$

- Non-classical information pattern
- Hard to find optimal solutions in general
- Special cases lead to tractable problems



Cf., Witsenhausen, Hu & Chu, Varaiya & Walrand, Borkar, Mitter & Tatikonda, Rotkowitz etc

Example

Plant

$$x_{k+1} = x_k + u_k + w_k, \quad x_0 = 2, \quad Ew_k^2 = 0.7^2$$

Certainty equivalent controller

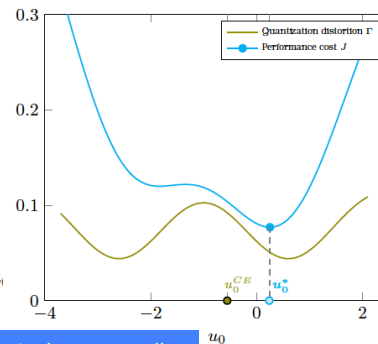
$$u_k^{\text{CE}} = -K_k^{\text{CE}} \left(E[x_k | \{y_k\}_0^k, \{u_k\}_0^{k-1}] + E[w_k | \{y_k\}_0^k, \{u_k\}_0^{k-1}] \right)$$

Event-generator encodes state as

$$\xi(x_k) = \begin{cases} 1, & \text{if } x_k \in (\infty, -\theta) \\ 2, & \text{if } x_k \in (-\theta, \theta) \\ 3, & \text{if } x_k \in (\theta, \infty) \end{cases}$$

Cost for time-horizon $N = 1$

$$J(u_0) = \sigma_w^2 + qu_0^2 + \left(p + \frac{qa^2}{q+1} \right) \mathbb{E}[x_1^2 | x_0, u_0]$$

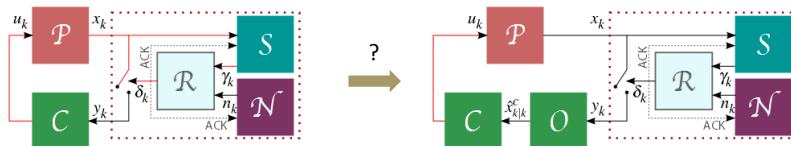


Optimal performance is not obtained by a certainty equivalent controller

Rabi et al, 2015

Condition for Certainty Equivalence

Corollary: The optimal controller for the system $\{\mathcal{P}, S(f), C(g)\}$, with respect to the cost J is certainty equivalent if the scheduling decisions are not a function of the applied controls.

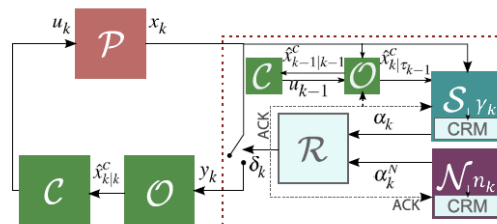


Certainty equivalence achieved at the cost of optimality

Bar-Shalom & Tse, 1974; Ramesh et al., 2011

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Architecture with Certainty Equivalent Controller



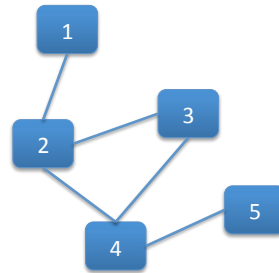
Ramesh et al., 2012, 2013

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Distributed Event-Based Control

- How to implement event-based control over a distributed system?
- Local control and communication, but global objective



Approach: Consider a prototype distributed control problem and study it under event-based communication and control

Average Consensus Problem

Multi-agent system model

- Group of N agents

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

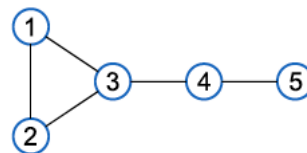
- Communication graph G
 A : undirected, connected

Adjacency matrix A with $a_{ij} = 1$ if agents i and j adjacent, otherwise $a_{ij} = 0$

Degree matrix D is the diagonal matrix with elements equal to the cardinality of the neighbor sets N_i

Objective: Average consensus

$$x_i(t) \xrightarrow{t \rightarrow \infty} a = \frac{1}{N} \sum_{i=1}^N x_i(0)$$



Consensus protocol

$$u_i(t) = - \sum_{j \in N_i} (x_i(t) - x_j(t))$$

Closed-loop dynamics

$$\dot{x}(t) = -Lx(t)$$

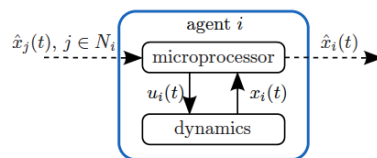
with Laplacian matrix $L = D - A$

Event-based implementation?

Olfati-Saber & Murray, 2004

Event-Based Average Consensus

Event-based scheduling of measurement broadcasts:



Event-based broadcasting

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

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

■ Consensus protocol

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

■ Measurement errors

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

■ Closed-loop

$$\dot{x}(t) = -L\hat{x}(t) = -L(x(t) + e(t))$$

■ Disagreement

$$\delta(t) = x(t) - a\mathbf{1}, \quad \mathbf{1}^T \delta(t) \equiv 0 \quad \text{Seyboth et al, 2013}$$

Trigger Function for Event-Based Control

Trigger mechanism: Define *trigger functions* $f_i(\cdot)$ and trigger when

$$f_i \left(t, x_i(t), \hat{x}_i(t), \bigcup_{j \in N_i} \hat{x}_j(t) \right) > 0$$

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

Extends [Tabuada, 2007] single-agent trigger function to multi-agent systems

Find f_i such that

- $|x_i(t) - x_j(t)| \rightarrow 0, t \rightarrow \infty$
- no Zeno (no accumulation point in time)
- few inter-agent communications

Cf., Dimarogonas et al., De Persis et al., Donkers et al., Mazo & Tabuada, Wang & Lemmon, Garcia & Antsaklis, Guinaldo et al.

Seyboth et al, 2013

Event-Based Control with Constant Thresholds

$$\dot{x}(t) = u(t), \quad u(t) = -L\hat{x}(t) \quad (1)$$

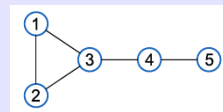
Theorem (constant thresholds)

Consider system (1) with undirected connected graph G . Suppose that

$$f_i(e_i(t)) = |e_i(t)| - c_0,$$

with $c_0 > 0$. Then, for all $x_0 \in \mathbb{R}^N$, the system does not exhibit Zeno behavior and for $t \rightarrow \infty$,

$$\|\delta(t)\| \leq \frac{\lambda_N(L)}{\lambda_2(L)} \sqrt{N} c_0.$$



Proof ideas:

- Analytical solution of disagreement dynamics yields

$$\|\delta(t)\| \leq e^{-\lambda_2(L)t} \|\delta(0)\| + \lambda_N(L) \int_0^t e^{-\lambda_2(L)(t-s)} \|e(s)\| ds$$

- Compute lower bound τ on the inter-event intervals Seyboth et al, 2013

Event-Based Control with Exponentially Decreasing Thresholds

$$\dot{x}(t) = u(t), \quad u(t) = -L\hat{x}(t) \quad (1)$$

Theorem (exponentially decreasing thresholds)

Consider system (1) with undirected connected graph G . Suppose that

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

with $c_1 > 0$ and $0 < \alpha < \lambda_2(L)$. Then, for all $x_0 \in \mathbb{R}^N$, the system does not exhibit Zeno behavior and as $t \rightarrow \infty$,

$$\|\delta(t)\| \rightarrow 0.$$

Remarks

- Asymptotic convergence: $|x_i(t) - x_j(t)| \rightarrow 0, t \rightarrow \infty$
- $\lambda_2(L)$ is the rate of convergence for continuous-time consensus, so threshold need to decrease slower

Seyboth et al, 2013

Event-Based Control with Exponentially Decreasing Thresholds and Offset

$$\dot{x}(t) = u(t), \quad u(t) = -L\hat{x}(t) \quad (1)$$

Theorem (exponentially decreasing thresholds with offset)

Consider system (1) with undirected connected graph G . Suppose that

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

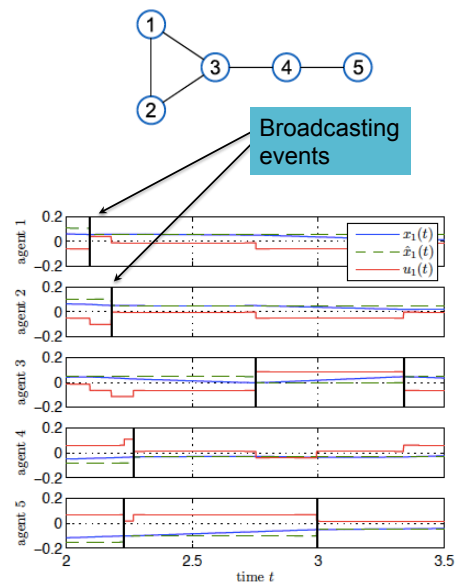
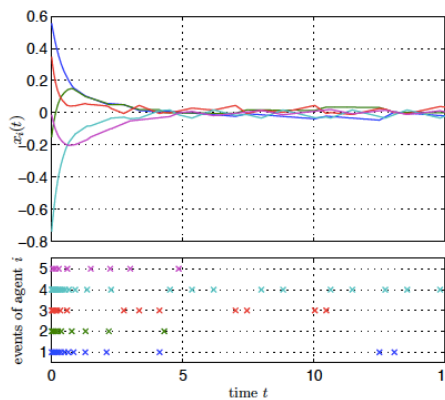
with $c_0, c_1 \geq 0$, at least one positive, and $0 < \alpha < \lambda_2(L)$. Then, for all $x_0 \in \mathbb{R}^N$, the system does not exhibit Zeno behavior and for $t \rightarrow \infty$,

$$\|\delta(t)\| \leq \frac{\lambda_N(L)}{\lambda_2(L)} \sqrt{N} c_0.$$

Seyboth et al, 2013

Example

$$\begin{aligned} \dot{x}(t) &= -L\hat{x}(t) \\ |e_i(t)| &\leq c_0 + c_1 e^{-\alpha t} \\ |e_i(t)| &\leq 0.050 \end{aligned}$$

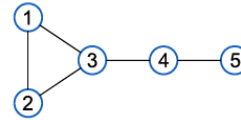


Seyboth et al, 2013

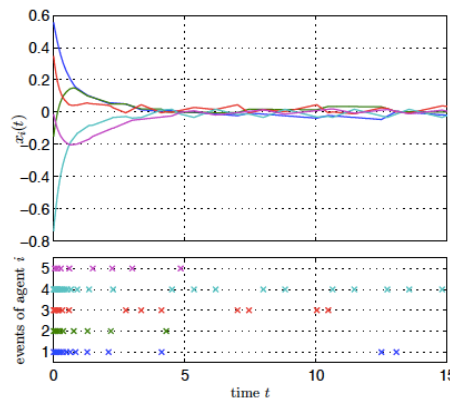
Example: Threshold Tuning

$$\dot{x}(t) = -L\hat{x}(t)$$

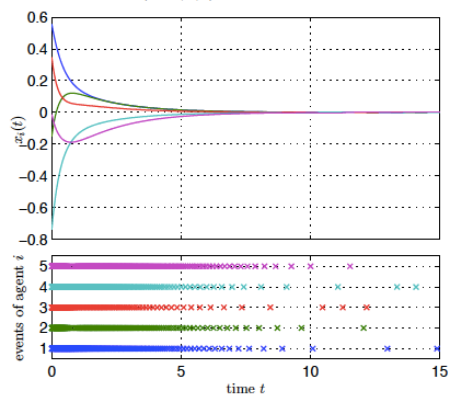
$$|e_i(t)| \leq c_0 + c_1 e^{-\alpha t}$$



$$|e_i(t)| \leq 0.050$$



$$|e_i(t)| \leq 0.001$$

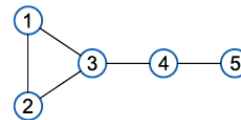


Seyboth et al, 2013

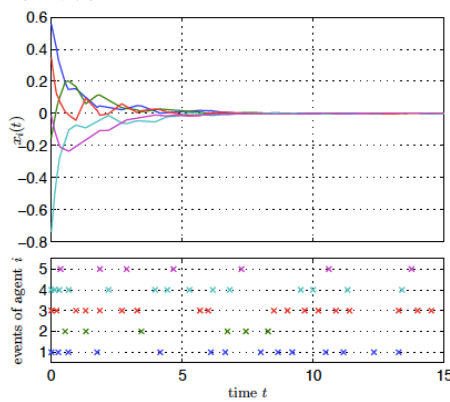
Example: Threshold Tuning

$$\dot{x}(t) = -L\hat{x}(t)$$

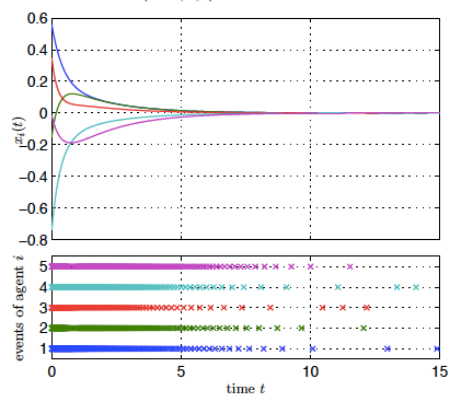
$$|e_i(t)| \leq c_0 + c_1 e^{-\alpha t}$$



$$|e_i(t)| \leq 0.001 + 0.249 e^{-0.9\lambda_2(L)t}$$



$$|e_i(t)| \leq 0.001$$

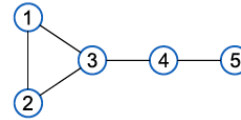


Seyboth et al, 2013

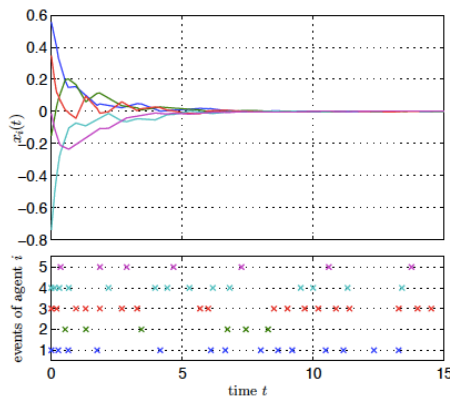
Example: Threshold Tuning

$$\dot{x}(t) = -L\hat{x}(t)$$

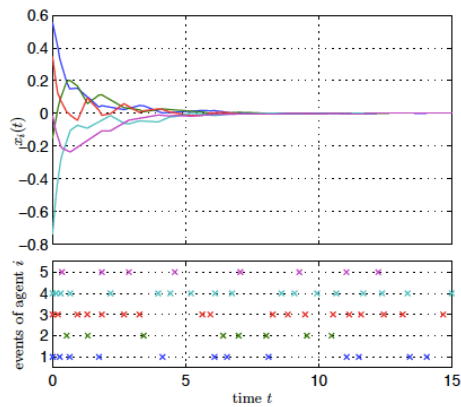
$$|e_i(t)| \leq c_0 + c_1 e^{-\alpha t}$$



$$|e_i(t)| \leq 0.001 + 0.249 e^{-0.9\lambda_2(L)t}$$

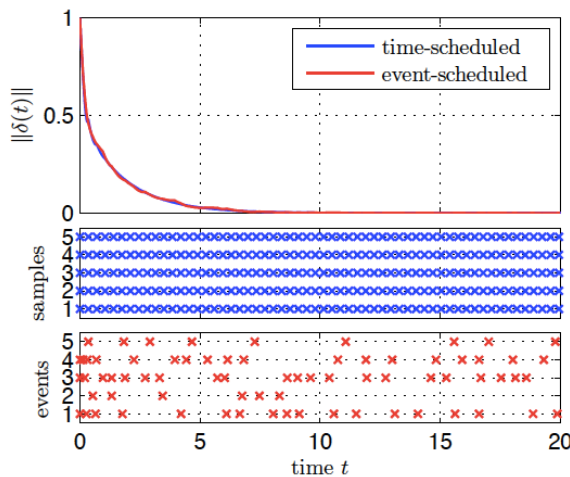


$$|e_i(t)| \leq 0.250 e^{-0.9\lambda_2(L)t}$$

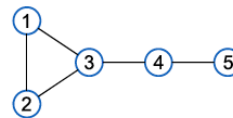


Seyboth et al, 2013

Example: Event- vs Time-Triggered Sampling



Graph:



Sampling periods:

■ Time-scheduling:

$$\tau_s = 0.350$$

$$\tau_{max} = 0.480$$

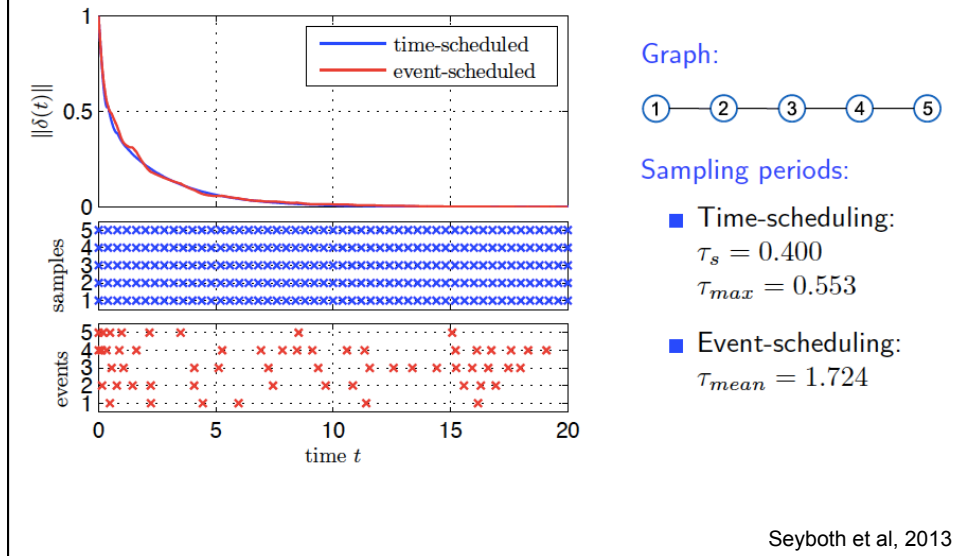
■ Event-scheduling:

$$\tau_{mean} = 1.389$$

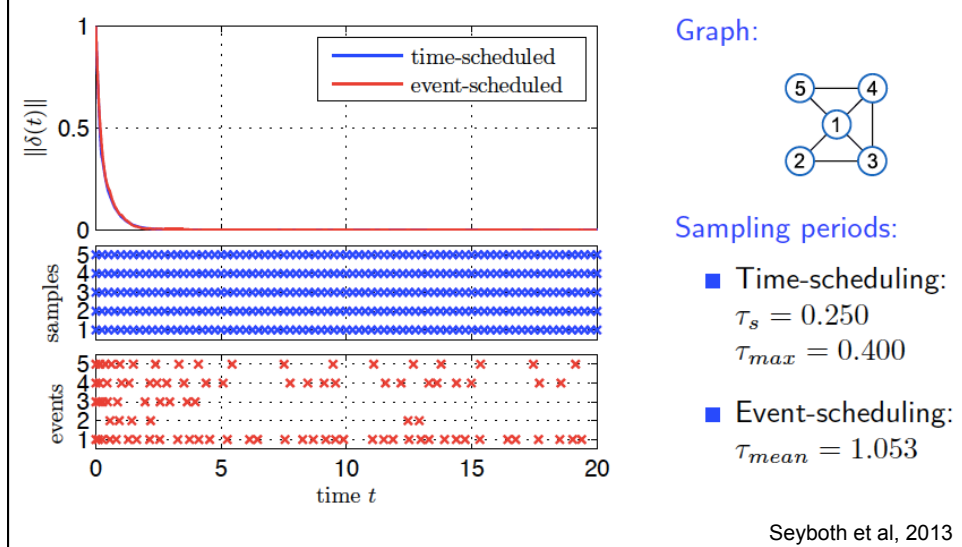
 τ_{max} : largest stabilizing sampling period, see [G. Xie et al., ACC2009](#)

Seyboth et al, 2013

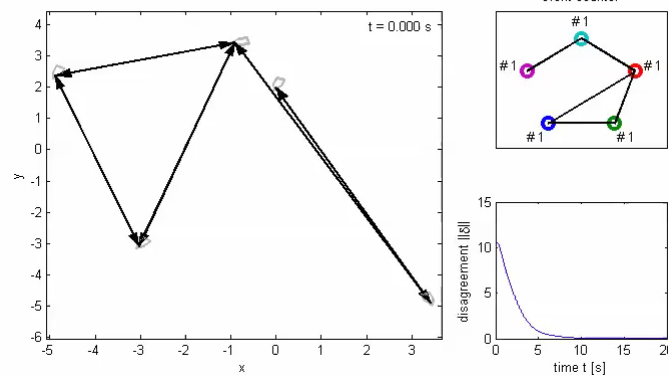
Example: Event- vs time-triggered sampling



Example: Event- vs time-triggered sampling



Event-Based Formation Control

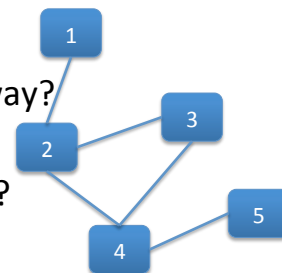


- Non-holonomic mobile robots under feedback linearization
- Event-based communication based on threshold for double-integrator network

Seyboth et al, 2013

Extensions

- How to **estimate** $\lambda_2(L)$ in a distributed way?
 - Aragues et al., 2014
- How to handle **general** agent **dynamics**?
 - Guinaldo et al. 2013
- How to handle network **delays** and packet **losses**?
 - Guinaldo et al., 2014
- **Pinning** (leader-follower) control and switching networks
 - Adaldo et al., 2015
- Event-triggered **pulse width modulation**
 - Meng et al., 2015
- Event-triggered **cloud access**
 - Adaldo et al., 2015

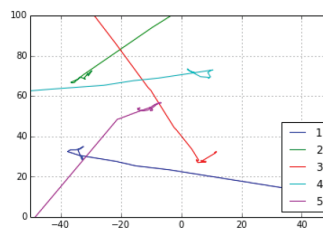
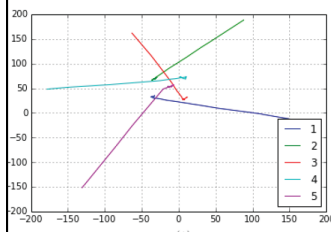
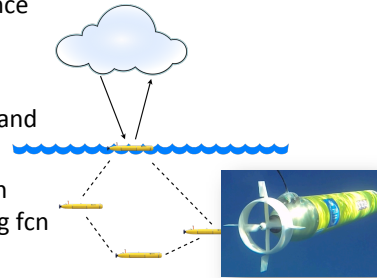


Event-triggered Cloud Access

- Agent dynamics with unknown drift disturbance

$$\dot{x}_i(t) = u_i(t) + \omega_i(t), \quad i = 1, \dots, N,$$

- Agents exchange state, control, disturbance, and timing data through a shared data base
- Schedule next data base access time based on dynamic estimates and event-based triggering fcn



Data base access times

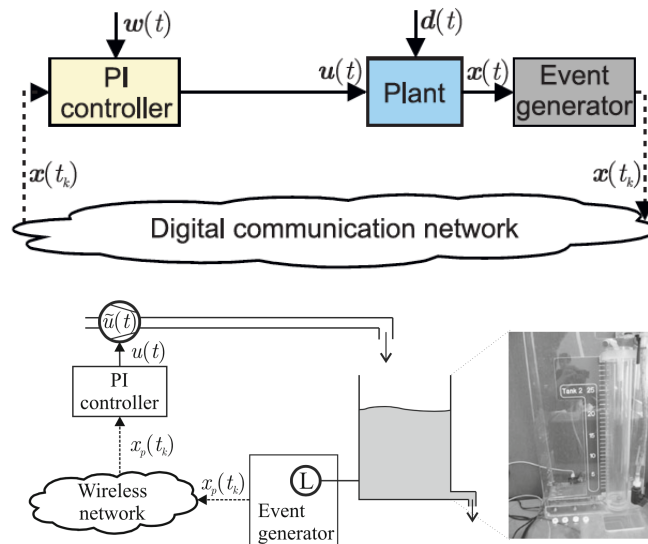
k	t _{1,k}	t _{2,k}	t _{3,k}	t _{4,k}	t _{5,k}
0	0.00	0.00	0.00	0.00	0.00
1	5.01	6.21	7.41	8.51	10.11
2	12.72	14.72	16.72	18.81	21.31
3	23.32	25.82	28.02	30.41	32.61
4	34.92	37.23	39.63	41.92	44.22
5	46.53	48.84			

Adaldo et al., 2015

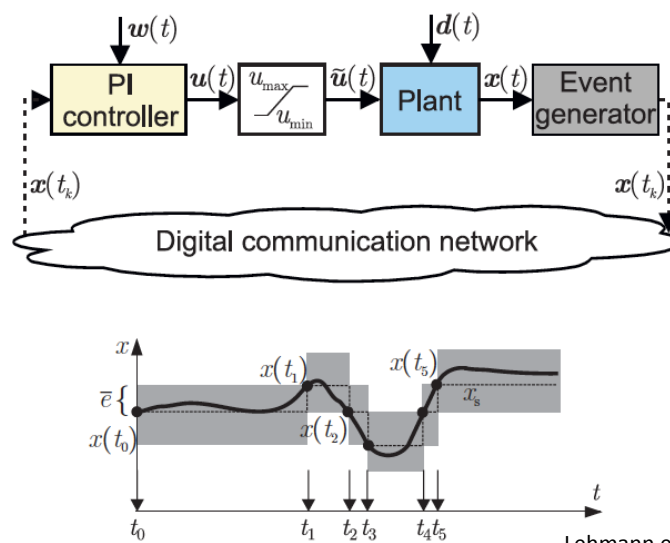
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Event-Based Wireless PI Control

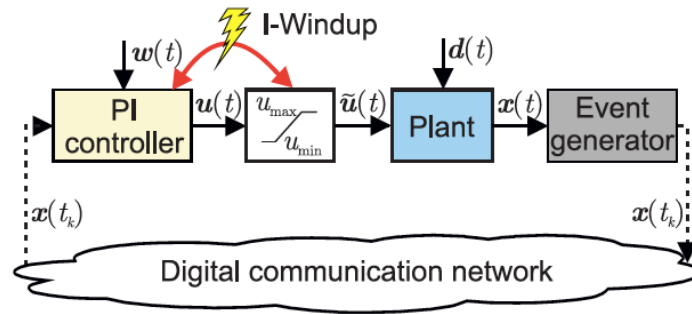


Event-Based PI Control with Saturation



Lehmann et al., 2012

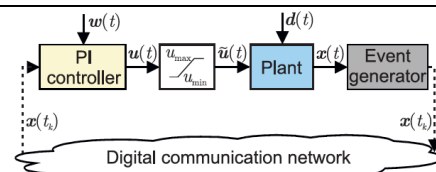
Event-Based PI Control with Saturation



► Industrial applications are generally affected by actuator limitations.

1. Does actuator saturation affect event-triggered PI control?
2. Under what conditions can we guarantee stability?
3. How to overcome potential effects of actuator saturation?

Example



► Plant:

$$\begin{aligned}\dot{x}(t) &= 0.1x(t) + \tilde{u}(t) + 0.1d(t), \quad x(0) = 0 \\ y(t) &= x(t)\end{aligned}$$

► Exogenous signals:

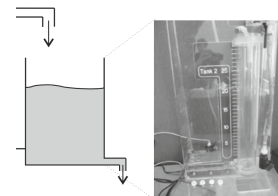
$$\begin{aligned}w(t) &= \bar{w} = 1.5 \\ d(t) &= \bar{d} = 0.1\end{aligned}$$

► Actuator saturation:

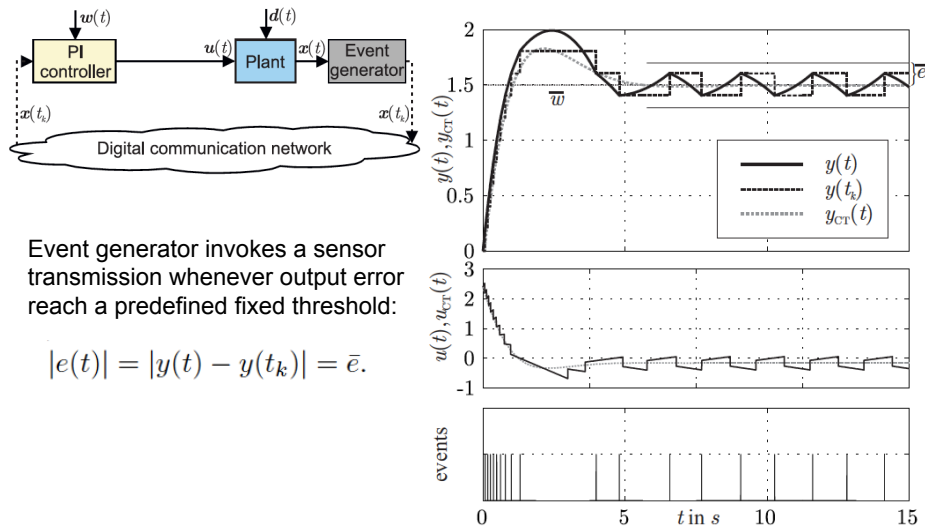
$$\tilde{u}(t) = \begin{cases} 0.4, & \text{for } u(t) > 0.4; \\ u(t), & \text{for } -0.4 \leq u(t) \leq 0.4; \\ -0.4, & \text{for } u(t) < -0.4; \end{cases}$$

► PI controller

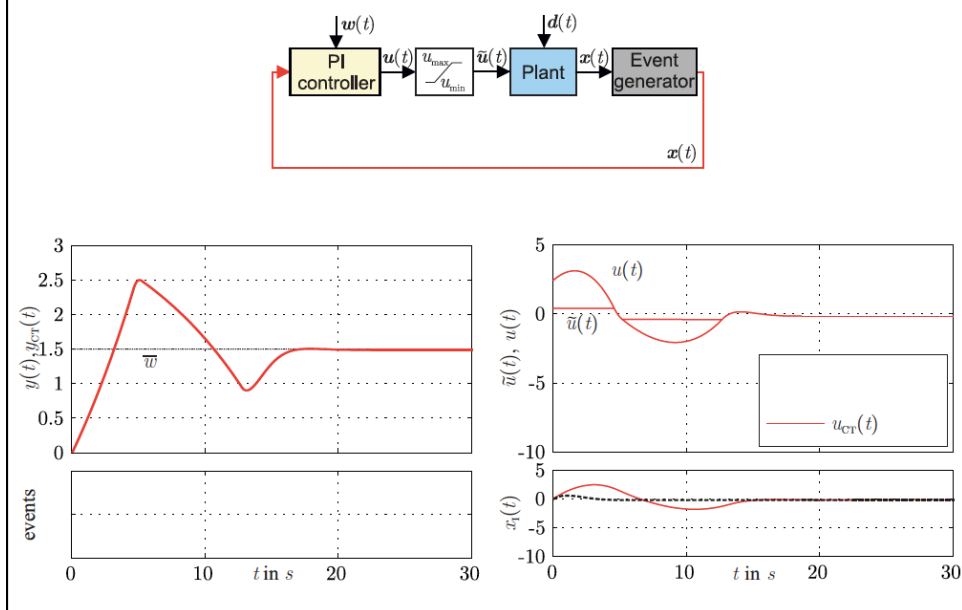
$$\begin{aligned}\dot{x}_I(t) &= y(t) - w(t), \quad x_I(0) = 0 \\ u(t) &= -x_I(t) - 1.6(y(t) - w(t))\end{aligned}$$



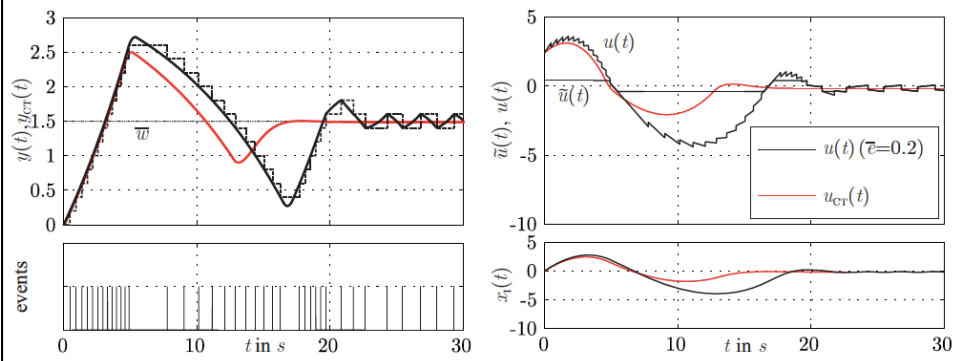
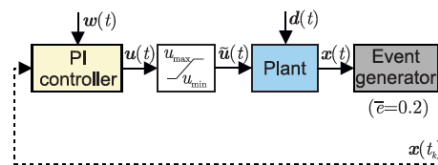
Example: Without Saturation



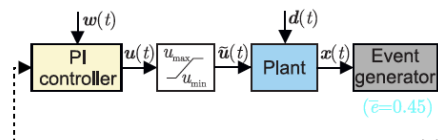
Motivating Example



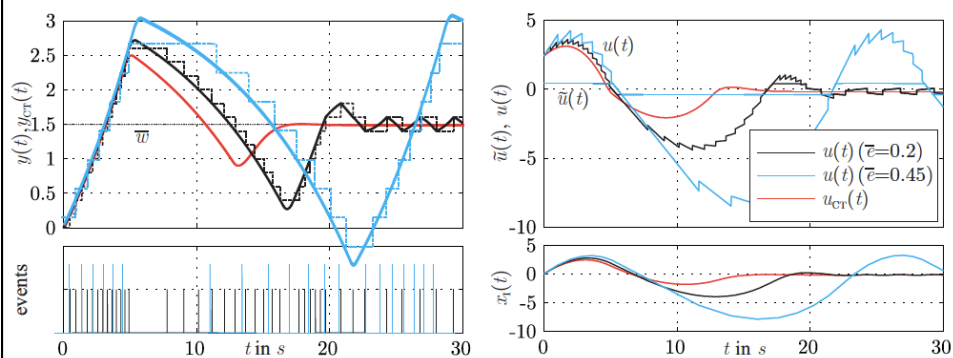
Motivating Example



Motivating Example



Need to take saturation and wind-up into account when designing event-based control systems



Mathematical Model

► Plant:

$$\dot{x}(t) = Ax(t) + B\tilde{u}(t) + Ed(t), \quad x(0) = x_0$$

$$\tilde{u}(t) = \text{sat}(u(t))$$

$$\text{sat}(u_i(t)) = \begin{cases} u_0, & \text{for } u_i(t) > u_0 \\ u_i(t), & \text{for } -u_0 \leq u_i(t) \leq u_0 \\ -u_0, & \text{for } u_i(t) < -u_0 \end{cases} \quad \forall i \in \{1, 2, \dots, m\}$$

► Event generator: $\|x(t) - x(t_k)\| = \bar{e}$

► PI controller:

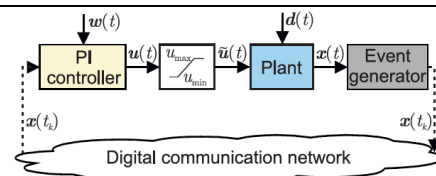
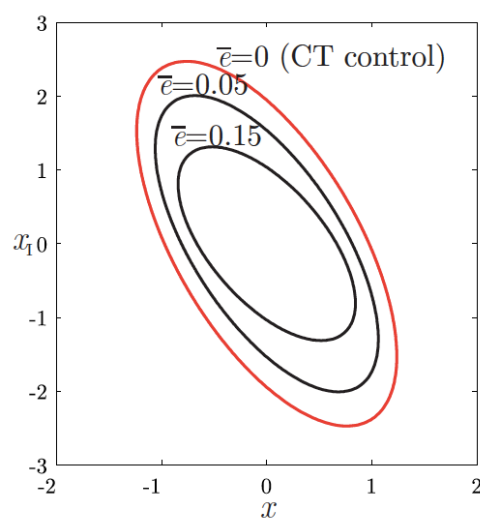
$$\dot{x}_I(t) = x(t) - e(t) - w(t), \quad x_I(0) = x_0$$

$$u(t) = K_I x_I(t) + K_P(x(t) - e(t) - w(t))$$

► **State error:** $e(t) = x(t) - x(t_k)$

► For the sake of simplicity: $w(t) = d(t) = 0$

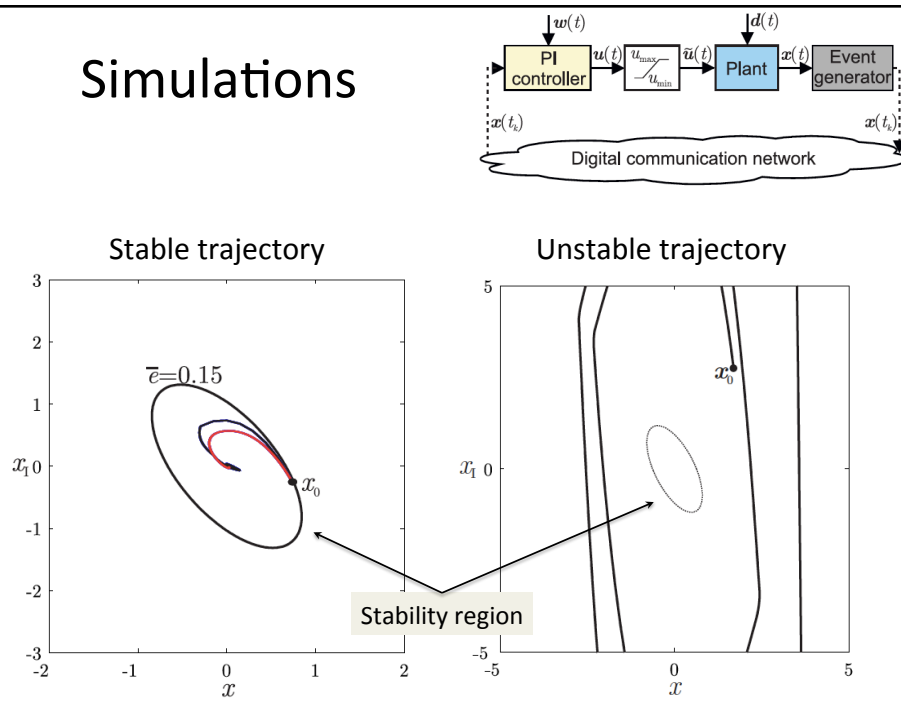
Stability Regions



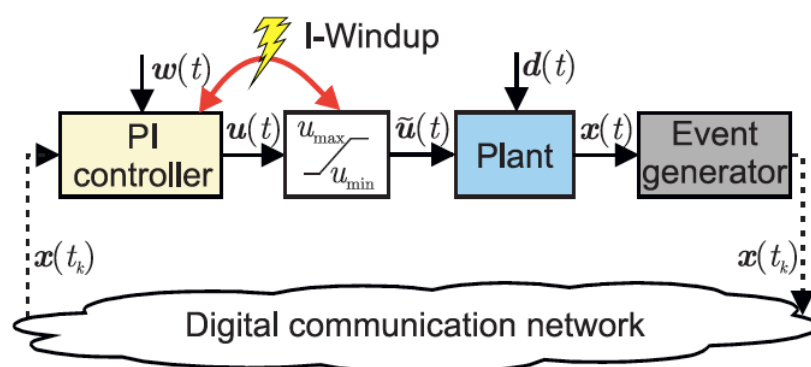
LMI condition to estimate region of stability

Lehmann et al., 2012

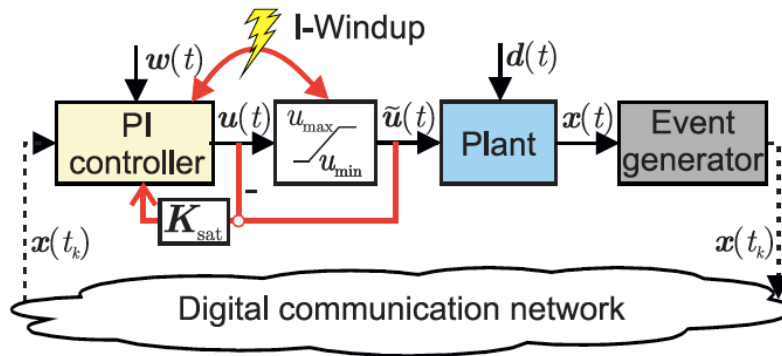
Simulations



Anti-Windup for Event-Based Control

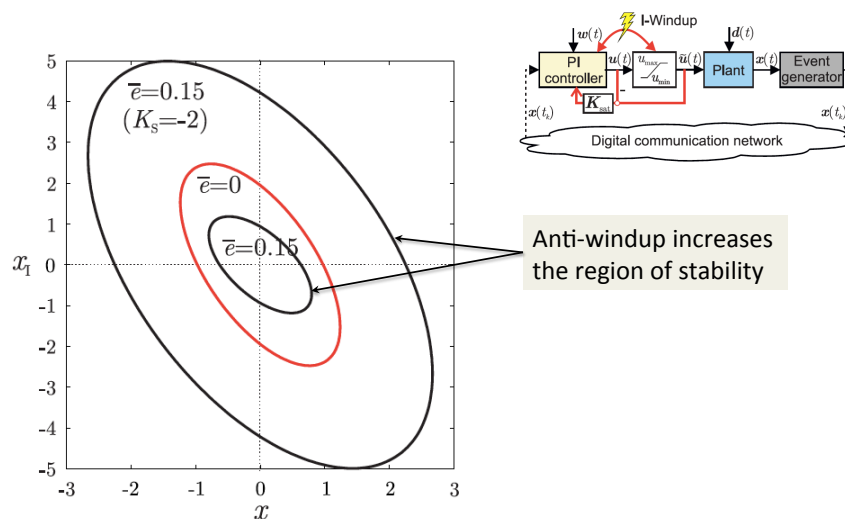


Anti-Windup for Event-Based Control

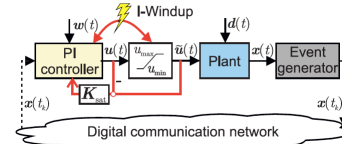
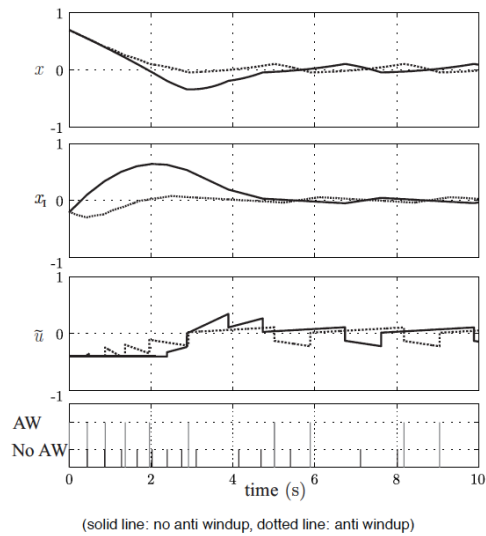


Cf., anti-windup for conventional control systems [Åström & Hägglund, 1995]

Stability Regions with Anti-Windup

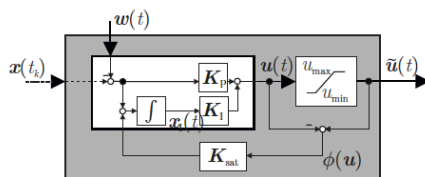


System Evolution with Anti-Windup

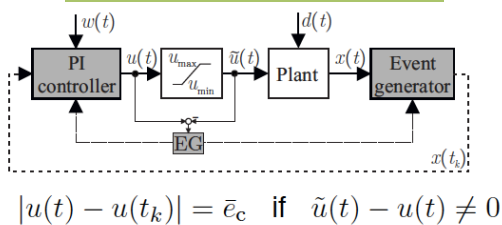


Anti-windup improves the system response

Event-Based Communication for Anti-Windup



Anti-windup event generated when actuator saturates (ETAW)

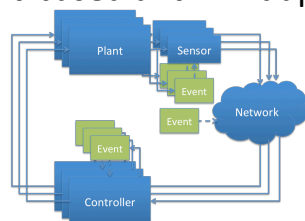


Outline

- Introduction
- Motivating applications
- Optimal event-based control
- Distributed event-based control
- Implementation aspects
- Conclusions

Conclusions

- **Event-based control** of multi-agent systems
- Hard to **jointly optimize** event condition and control law
- Certain **architectures** lead to strong results
- **Applications** in goods transportation, mobile robotics, and wireless automation
- Event-based **revisions** of classical control architectures: event-based anti-windup, feedforward, cascade control



<http://people.kth.se/~kallej>