



Wireless Event-Triggered Control

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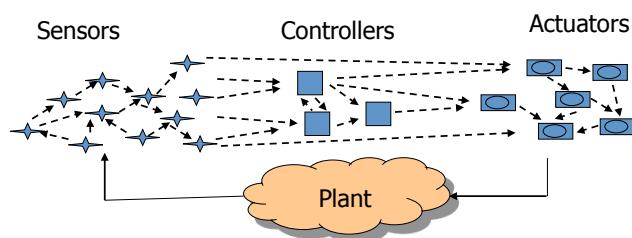
Based on joint work with Chithrupa Ramesh, José Araujo, Maben Rabi, Georg Seyboth,
 Henrik Sandberg, Carlo Fischione, Dimos Dimarogonas



Tutorial Session on Event-triggered and Self-triggered Control, IEEE CDC, Maui, 2012

Wireless control system

How to share common network resources while
 maintaining guaranteed closed-loop performance?



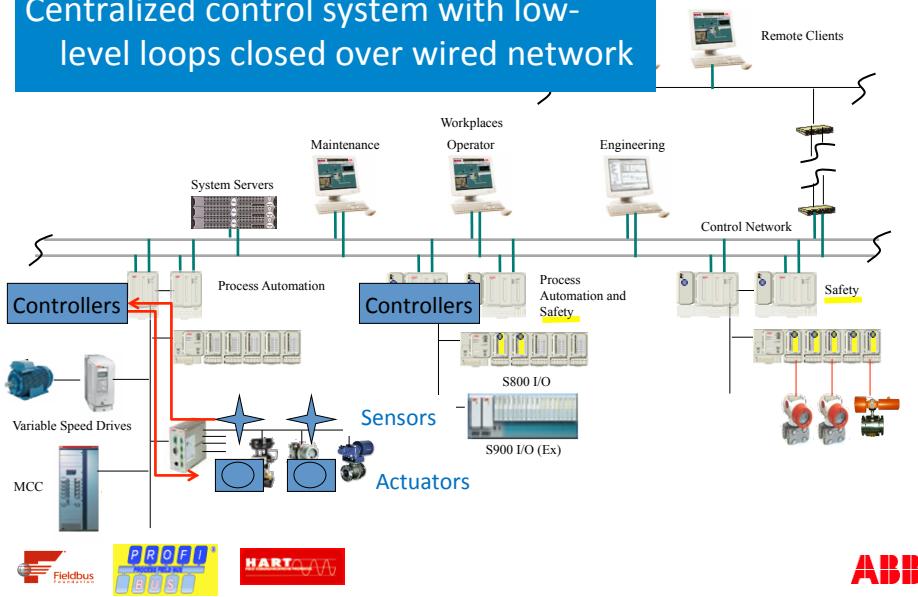
Idea: Utilize event- and self-triggered control to limit the use of network resources

Outline

- Motivating industrial applications
- Event-based scheduling for stochastic control
- Exploiting wireless network protocols
- Event-based control over lossy networks
- Extensions
- Conclusions

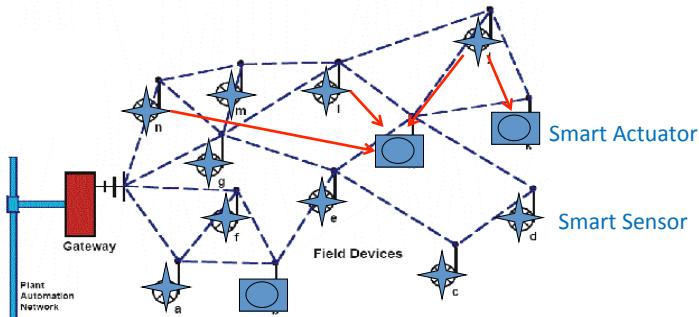
Today's industrial communication architecture

Centralized control system with low-level loops closed over wired network

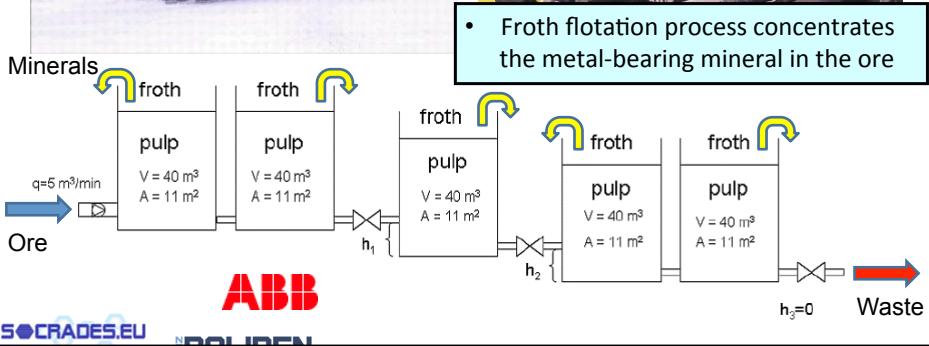


Towards wireless sensor and actuator network architecture

- Local control loops closed over **wireless multi-hop network**
- Potential for a dramatic change:
 - From fixed hierarchical centralized system to flexible distributed
 - Move intelligence from dedicated computers to sensors/actuators



Event-based control of froth flotation process



Wireless control of floatation process

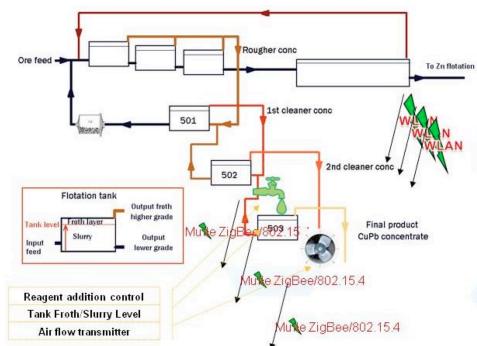
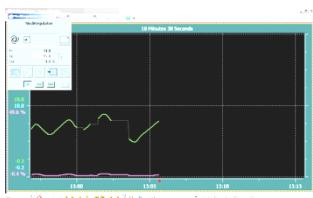
The Boliden plant



Existing wired communication system



Wireless communication for tank level control

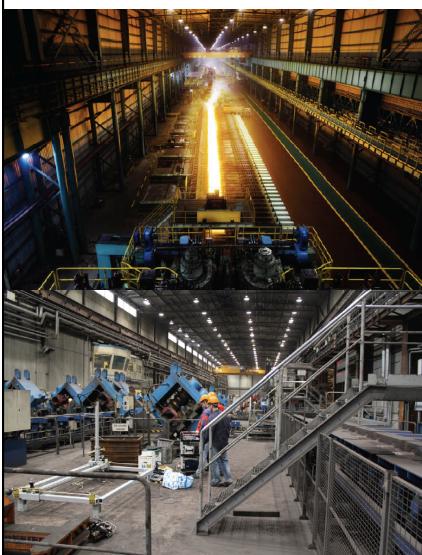


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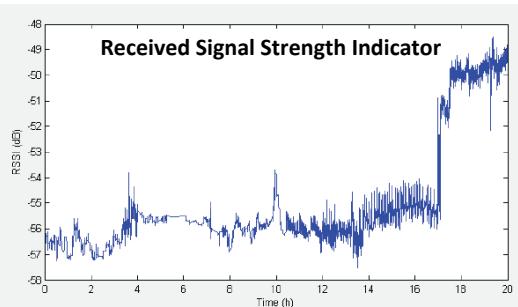
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Radio Channel Measurements in Industrial Environment



- Rolling mill at Sandvik in Sweden
 - Study of 2.45 GHz radio channel properties
 - Slow but substantial RSSI variations due to mobile machines





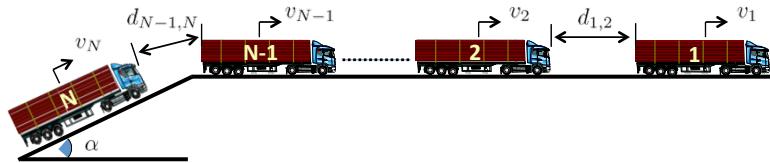
UPPSALA
UNIVERSITET

ABB

 KTH
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Ahlen et al, 2012

Event-based estimation in vehicle platooning



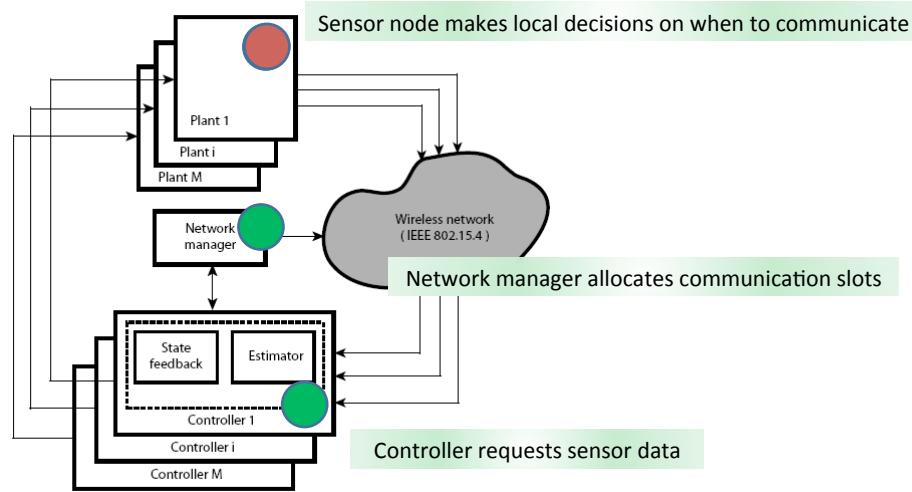
- Vehicles need **accurate estimates** of neighboring vehicles' states and actions
- Control performance is tightly coupled to how well data (position, velocity, breaking estimates) are communicated across the platoon
- Today's communication protocols are event-based (e.g., IEEE 801.11p)



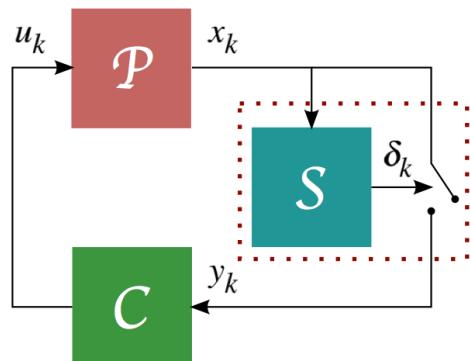
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Where to take medium access decisions?



Is there a separation between event-based scheduling-estimation-control?



Stochastic control formulation

Plant:

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

Scheduler:

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

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

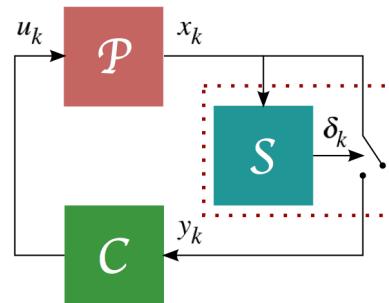
Controller:

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

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

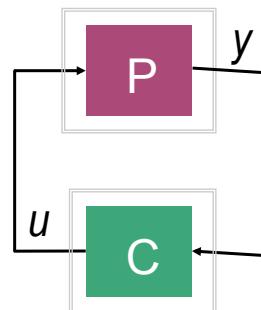
Cost criterion:

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



Certainty equivalence revisited

Definition Certainty equivalence holds if the closed-loop optimal controller has the same form as the deterministic optimal controller with x_k replaced by the estimate $\hat{x}_{k|k} = \mathbb{E}[x_k | \mathbb{I}_k^C]$.



Theorem [Bar-Shalom–Tse] Certainty equivalence holds if and only if $E[(x_k - E[x_k | I_k^c])^2 | I_k^c]$ is not a function of past controls $\{u\}_0^{k-1}$ (no dual effect).

Feldbaum, 1965; Åström, 1970; Bar-Shalom and Tse, 1974

Event-based scheduler

Plant:

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

Scheduler:

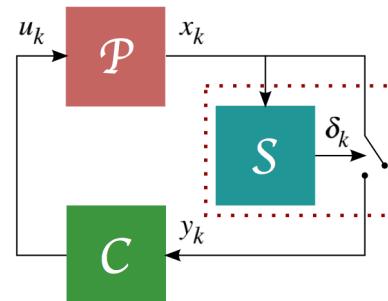
$$\delta_k = f_k(\mathbb{I}_k^S) \in \{0, 1\}$$

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

Controller:

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

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



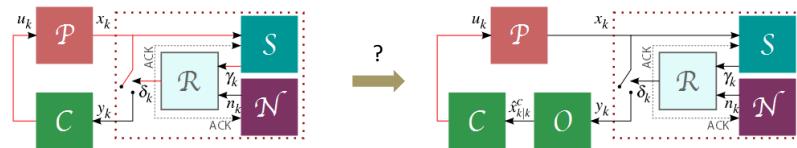
Corollary The control u_k for the optimal closed-loop system has a dual effect.

The separation principle does not hold for the optimal closed-loop system, so the design of the (event-based) scheduler, estimator, and controller is coupled

Ramesh et al., 2011

Conditions for Certainty Equivalence

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



Certainty equivalence achieved at the cost of optimality

Ramesh et al., 2011

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Event-based control architecture

- Plant \mathcal{P} :

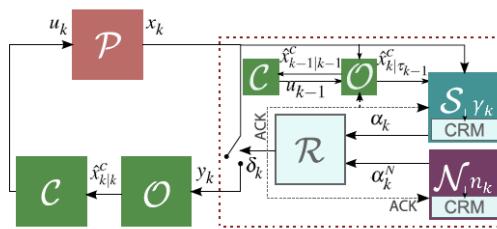
$$x_{k+1} = ax_k + bu_k + w_k$$
- CRM: $\mathbb{P}(\alpha_k=1|\gamma_k=1) = \mathbb{P}(\alpha_k^N=1|n_k=1) = p_\alpha$

$$\delta_k = \alpha_k(1 - \alpha_k^N)$$
- State-based Scheduler \mathcal{S} :

$$\gamma_k = \begin{cases} 1, & |x_k - \hat{x}_{k|\tau_{k-1}}|^2 > \epsilon_d, \\ 0, & \text{otherwise.} \end{cases}$$

$$\hat{x}_{k|\tau_{k-1}}^s = a\hat{x}_{k-1|\tau_{k-1}}^c + bu_{k-1}$$
- Observer \mathcal{O} : $y_k^{(j)} = \delta_k^{(j)} x_k^{(j)}$

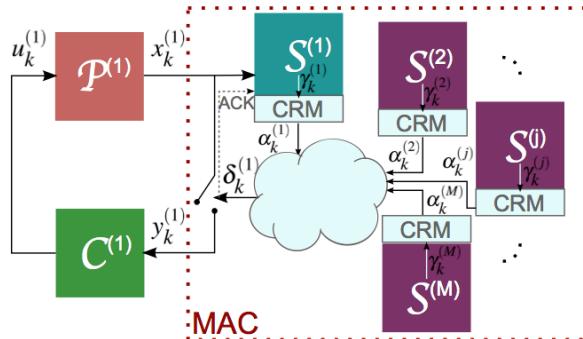
$$\hat{x}_{k|k}^c = \bar{\delta}_k(a\hat{x}_{k-1|k-1}^c + bu_{k-1}) + \delta_k x_k$$
- Controller \mathcal{C} : $u_k = -L\hat{x}_{k|k}^c$



Ramesh et al., CDC, 2012, ThC01.3

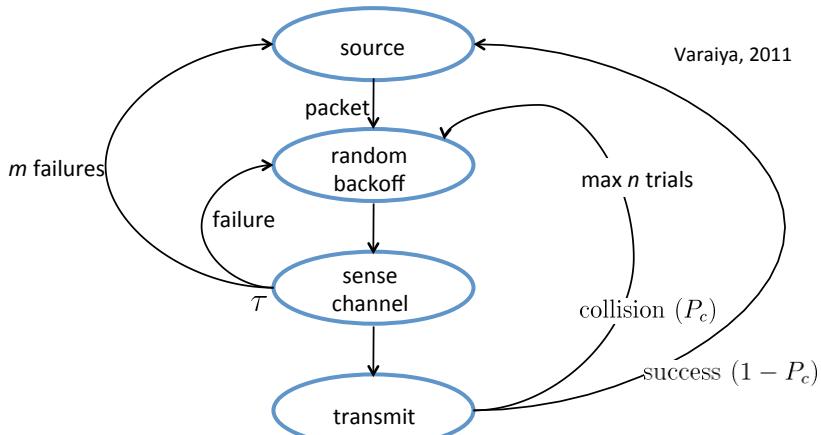
Integrating advanced contention resolution mechanisms

- Hard problem because of correlation between transmissions (and the plant states)
- Closed-loop analysis can still be done for classes of event-based schedulers and MAC's



Ramesh et al., CDC 2011

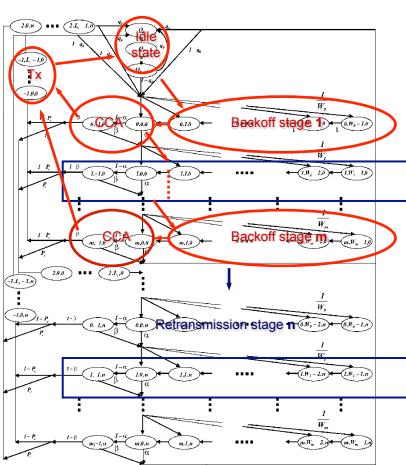
Contention resolution through CSMA/CA



- Every transmitting device executes this protocol
- For analysis, assume carrier sense events are independent [Bianchi, 2000]

CSMA/CA = Carrier Sense Multiple Access with Collision Avoidance

Detailed model of CSMA/CA in IEEE 802.15.4



- Markov state (s, c, r)
 - s: backoff stage
 - c: state of backoff counter
 - r: state of retransmission counter
- Model parameters
 - q_0 : traffic condition ($q_0=0$ saturated)
 - m_o, m, m_b, n : MAC parameters
- Computed characteristics
 - α : busy channel probability during CCA1
 - β : busy channel probability during CCA2
 - P_c : collision probability
- Validated in simulation and experiment
- Reduced-order models for control design
- Detailed model for numerical evaluations

Park, Di Marco, Soldati, Fischione, J, 2009

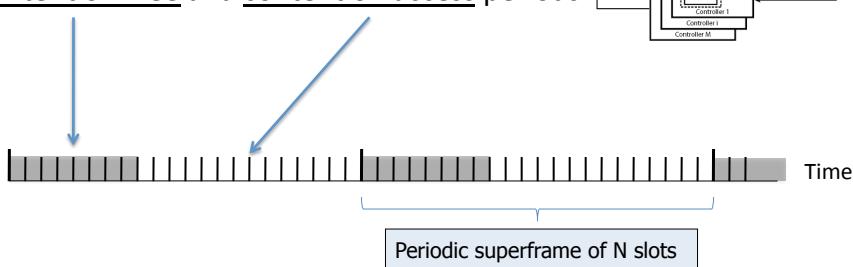
Cf., Bianchi, 2000; Pollin et al., 2006

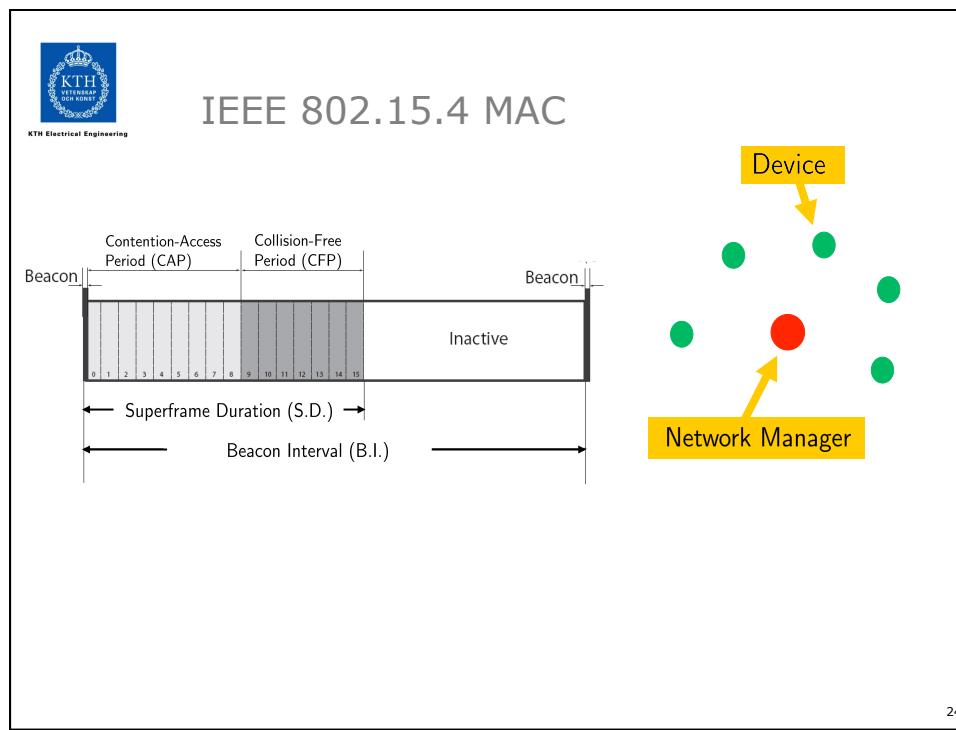
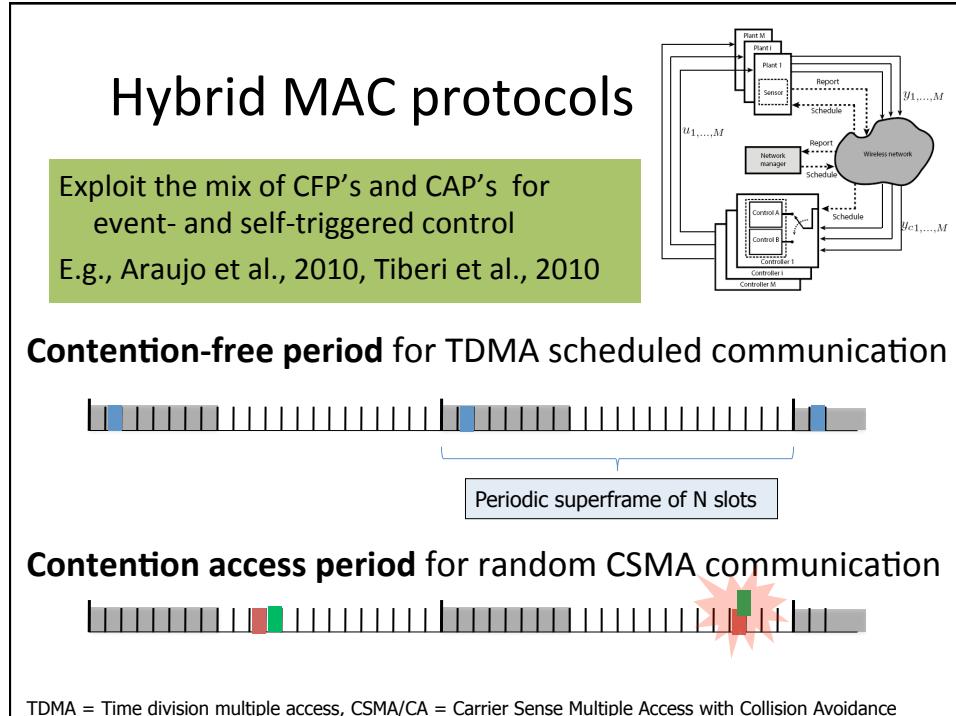
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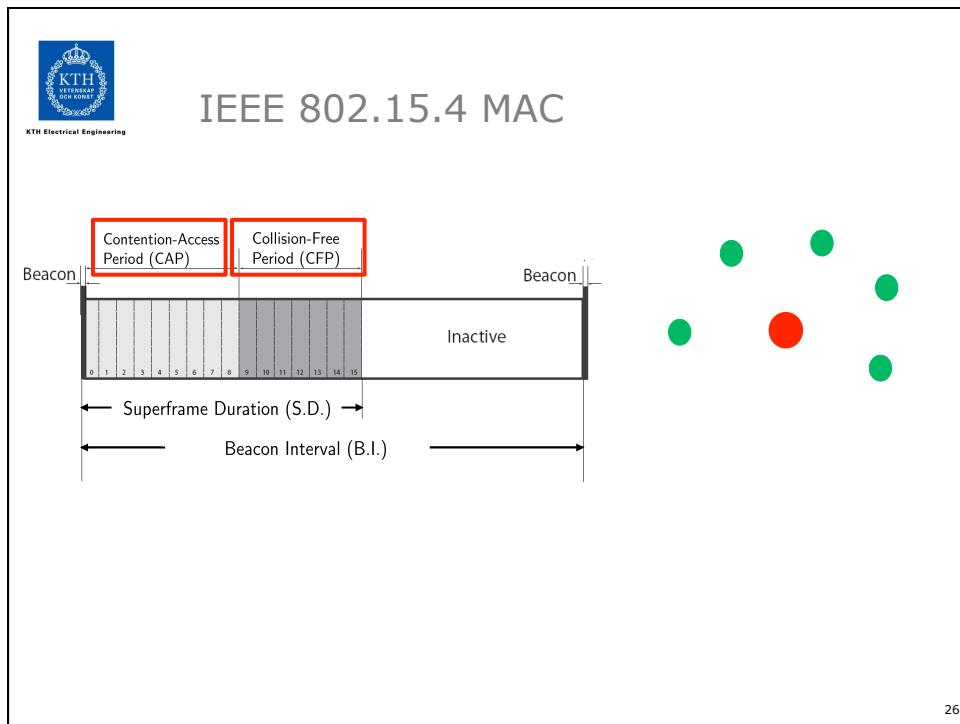
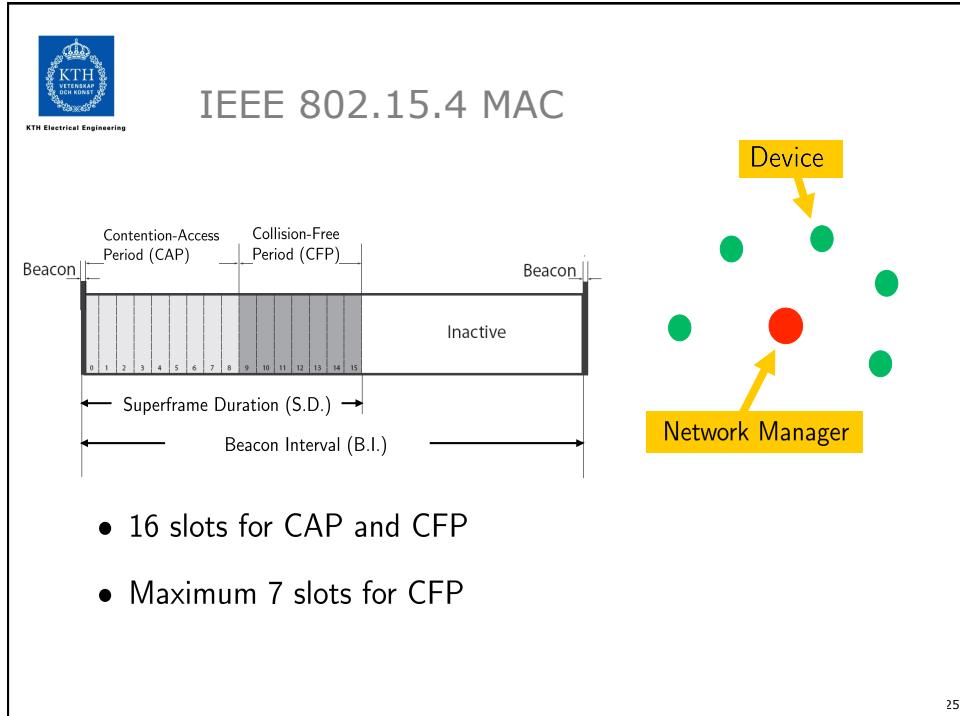
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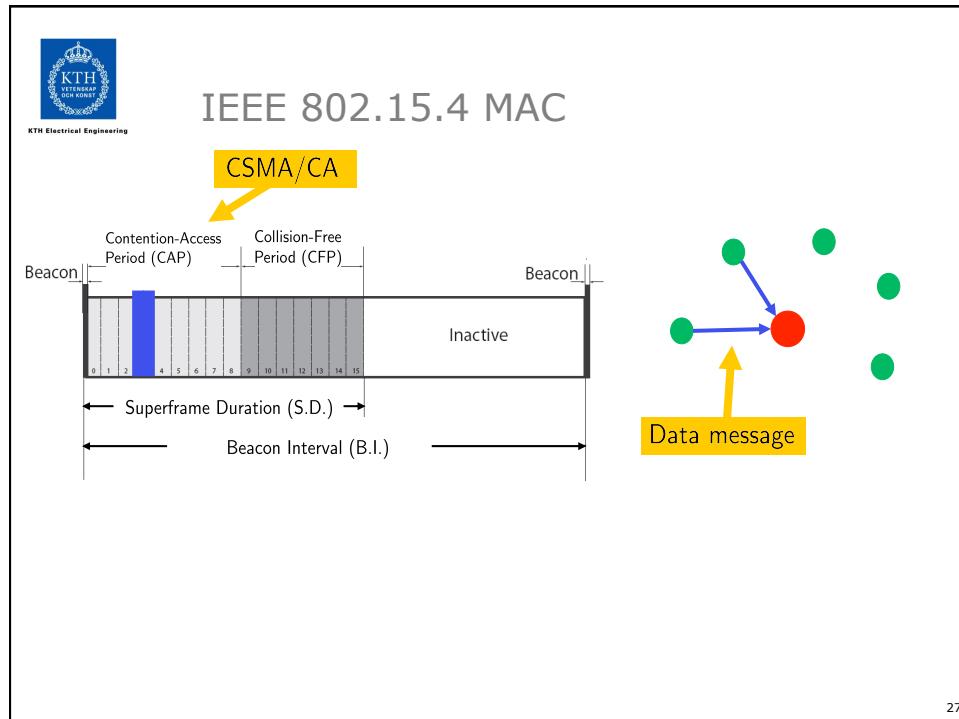
Slotted medium access

Many medium access protocols have slotted
contention-free and contention access periods

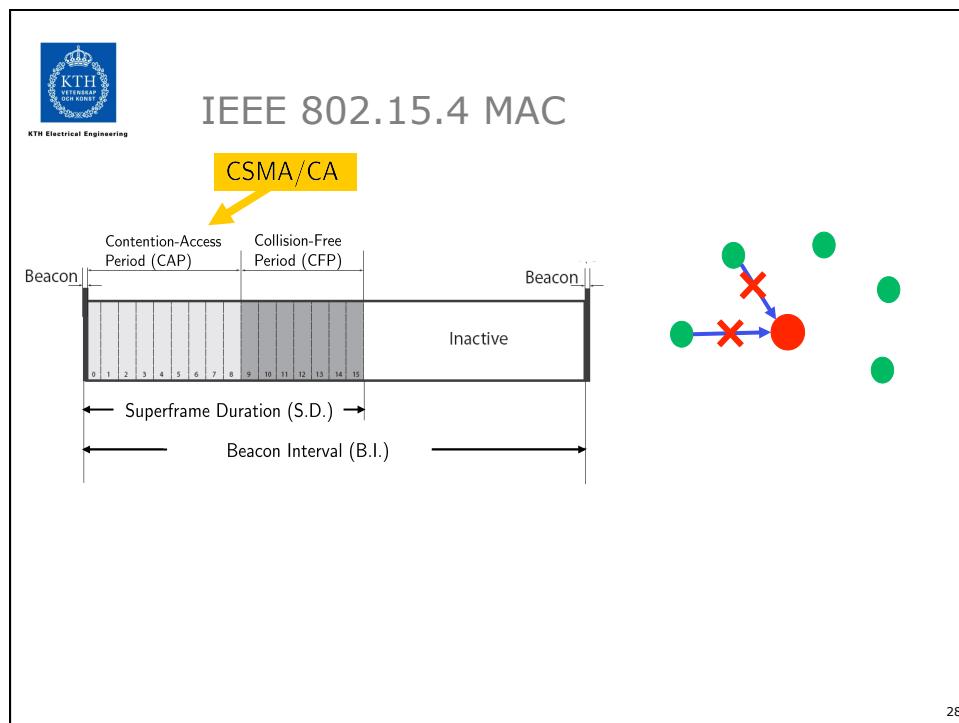




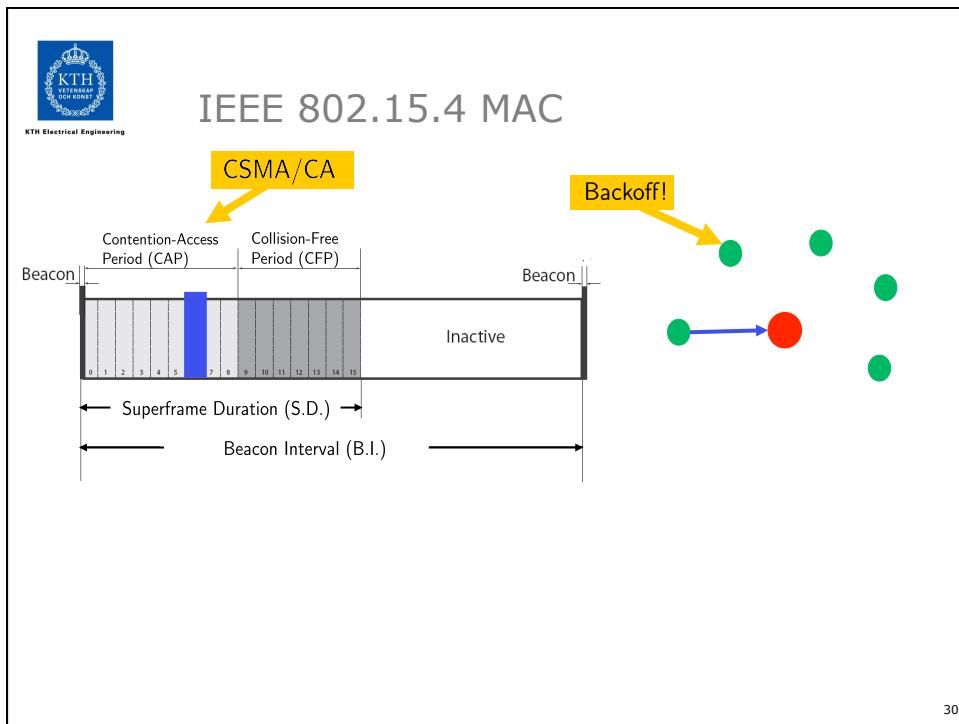
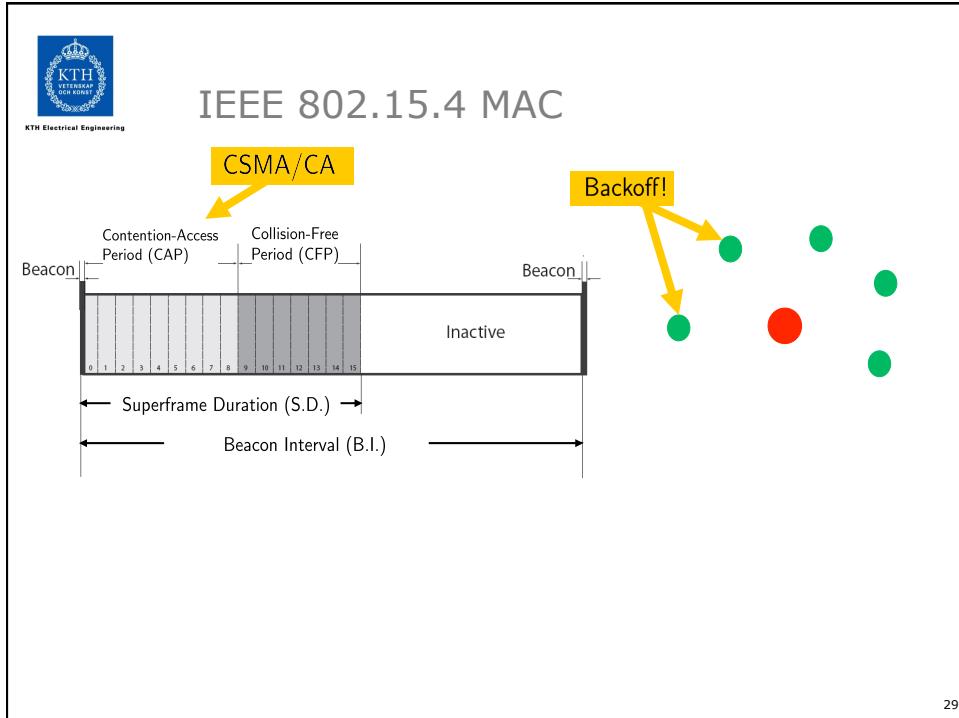


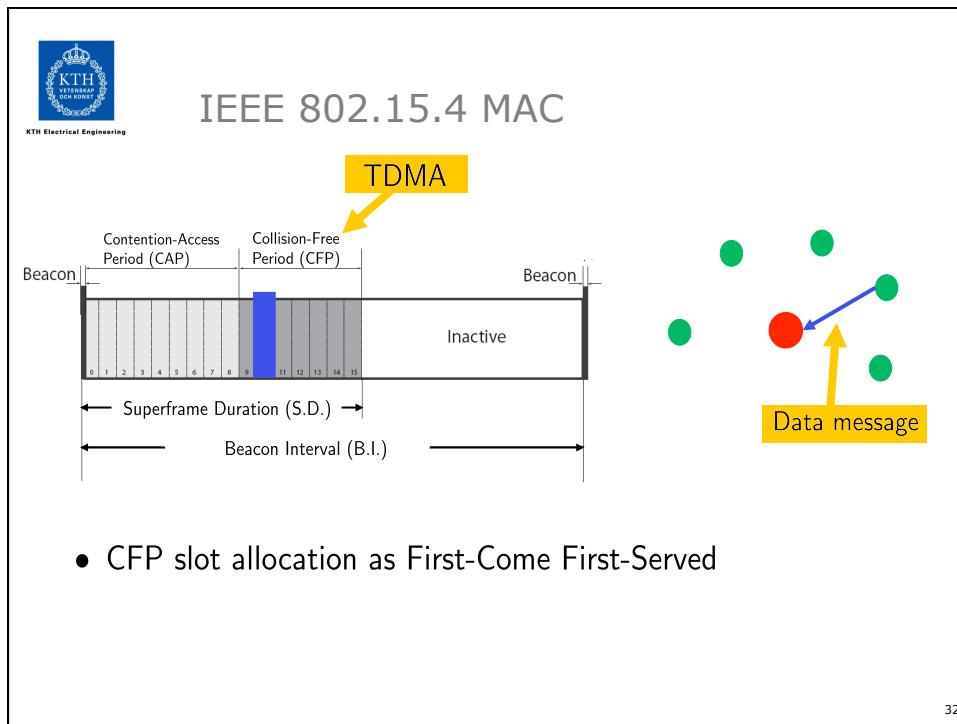
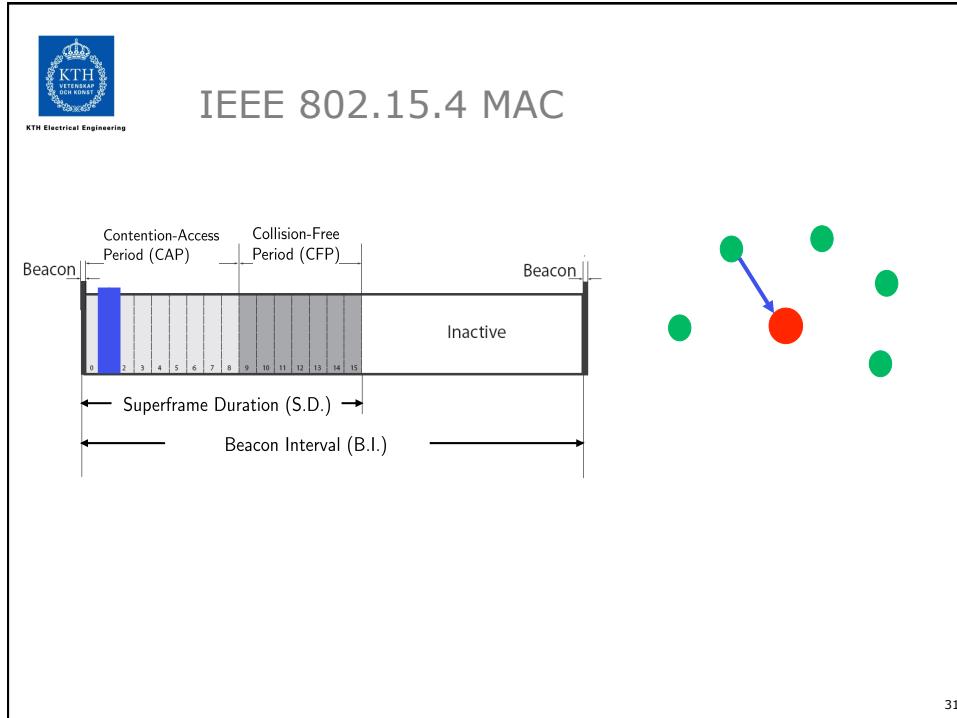


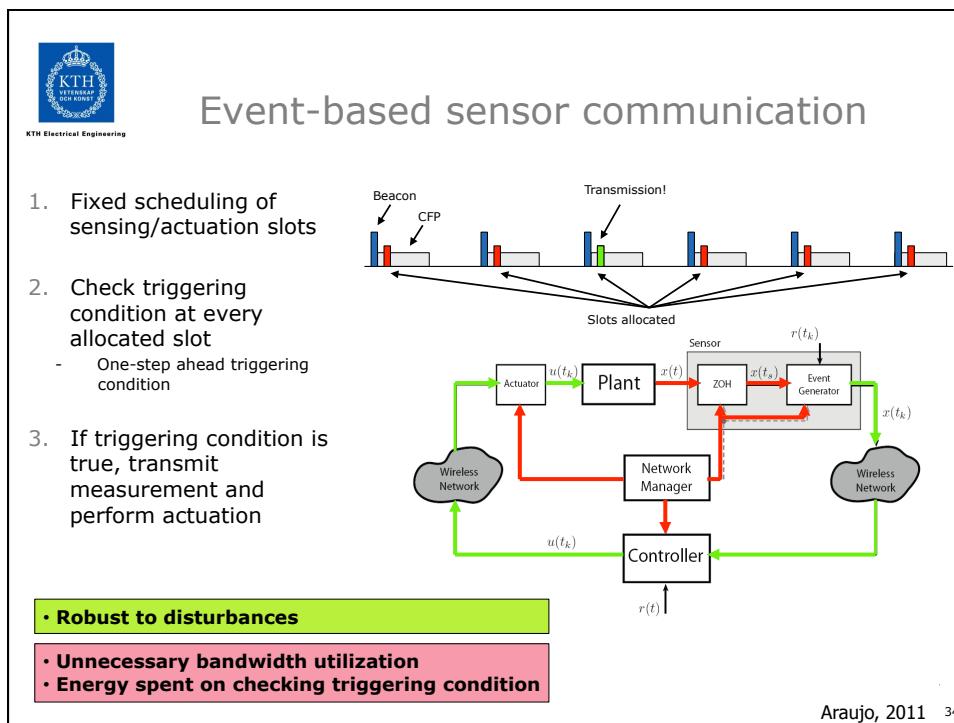
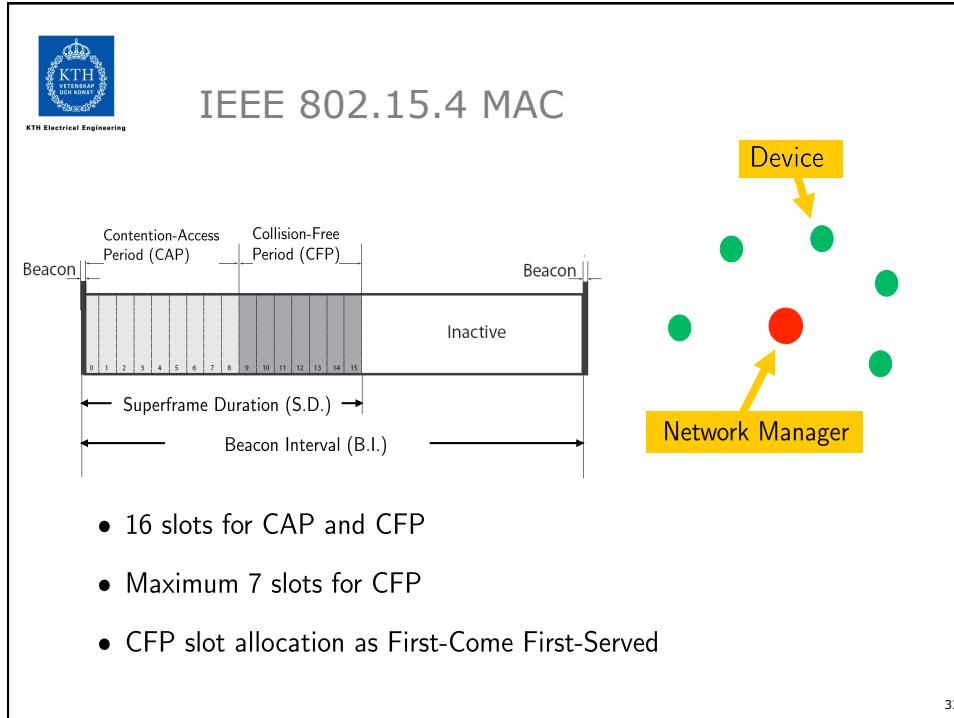
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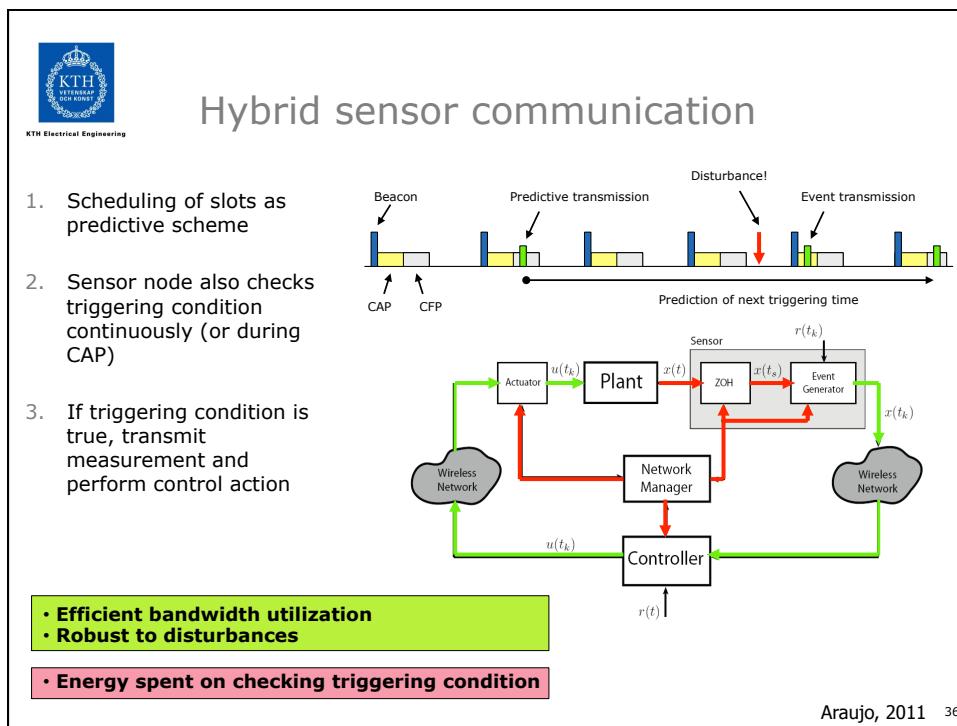
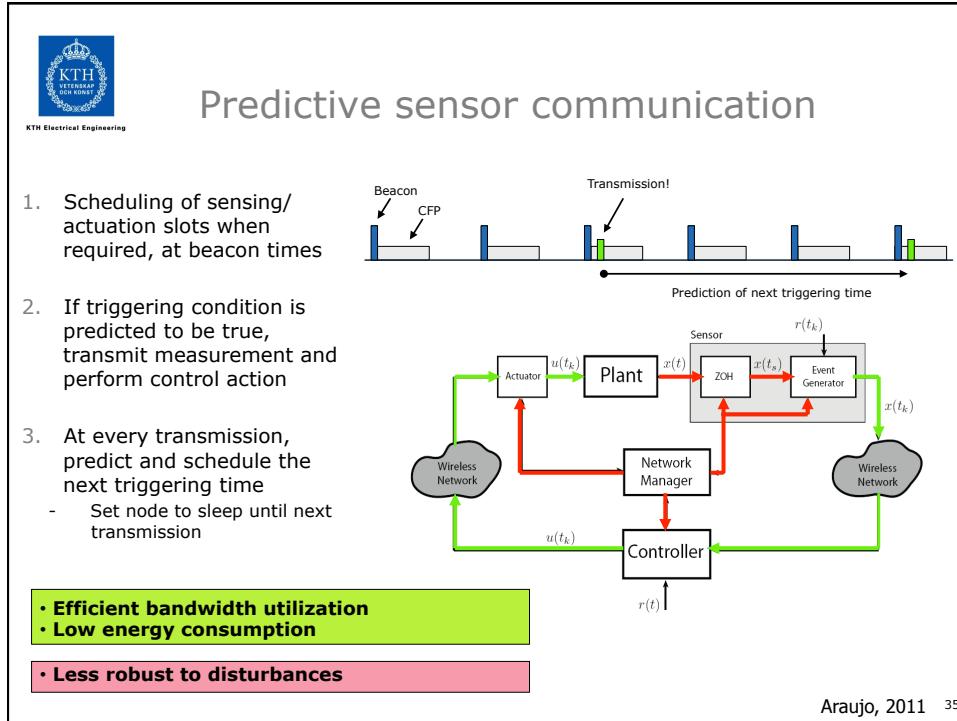


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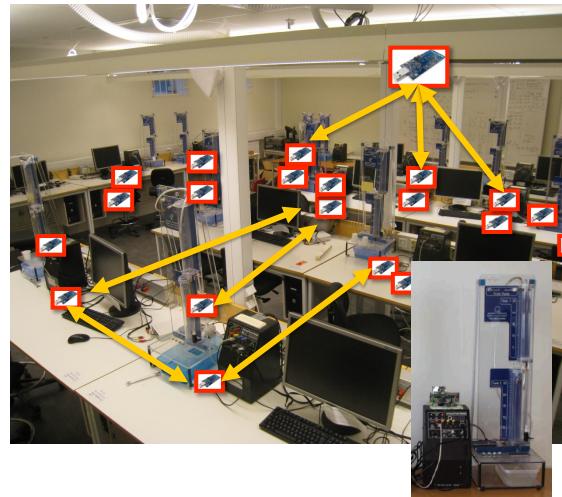
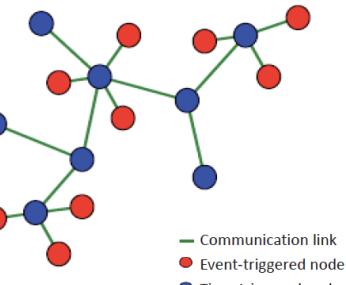






Multi-hop networks

- Routing decisions
- Time delays
- Hidden terminal problem



Outline

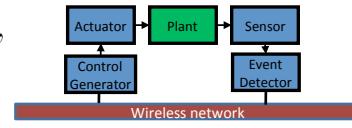
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Event-based impulse control

Plant $dx_t = dW_t + u_t dt, \quad x(0) = x_0,$

Sampling events $\mathcal{T} = \{\tau_0, \tau_1, \tau_2, \dots\},$

Impulse control $u_t = \sum_{n=0}^{\infty} x_{\tau_n} \delta(\tau_n)$



Average sampling rate $R_\tau = \limsup_{M \rightarrow \infty} \frac{1}{M} \mathbb{E} \left[\int_0^M \sum_{n=0}^{\infty} \mathbf{1}_{\{\tau_n \leq M\}} \delta(s - \tau_n) ds \right]$

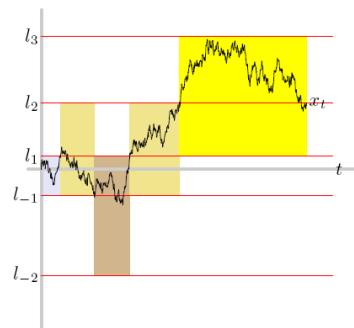
Average cost $J = \limsup_{M \rightarrow \infty} \frac{1}{M} \mathbb{E} \left[\int_0^M x_s^2 ds \right]$

Level-triggered control

Ordered set of levels $\mathcal{L} = \dots, l_{-2}, l_{-1}, l_0, l_1, l_2, \dots$ $l_0 = 0$

Multiple levels needed because we allow packet loss

Sampling instances $\tau = \inf \{ \tau \mid \tau > \tau_i, x_\tau \in \mathcal{L}, x_\tau \notin x_{\tau_i} \}$



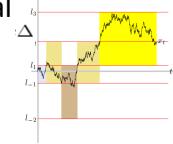
Level-triggered control

For Brownian motion, equidistant sampling is optimal

$$\mathcal{L}^* = \{k\Delta | k \in \mathbb{Z}\}$$

First exit time

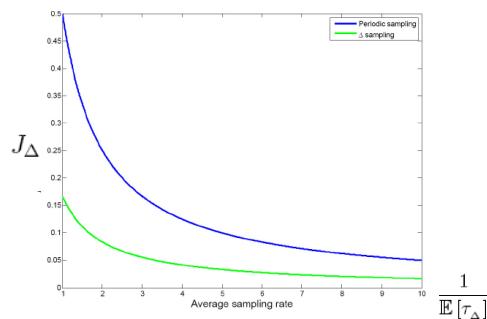
$$\tau_\Delta = \inf \{\tau | \tau \geq 0, x_\tau \notin (\xi - \Delta, \xi + \Delta), x_0 = \xi\}$$



Average sampling rate $R_\Delta = \frac{1}{\mathbb{E}[\tau_\Delta]} = \frac{1}{\Delta^2}$,

Average cost $J_\Delta = \frac{\mathbb{E}[\int_0^{\tau_\Delta} x_s^2 ds]}{\mathbb{E}[\tau_\Delta]} = \frac{\Delta^2}{6}$.

Comparison between **time-** and **event-based** control



$T = \Delta^2$ gives equal average sampling rate for periodic control and event-based control

Event-based impulse control is three times better than periodic

Åström & Bernhardsson, IFAC, 1999

What about the influence of communication losses?
Is event-based sampling still better?

Influence of i.i.d. packet loss

Times when packets are successfully received $\rho_i \in \{\tau_0 = 0, \tau_1, \tau_2, \dots\}$,

$$\{\rho_0 = 0, \rho_1, \rho_2, \dots\} . \quad \rho_i \geq \tau_i,$$

Average rate of packet reception

$$R_p = \limsup_{M \rightarrow \infty} \frac{1}{M} \mathbb{E} \left[\int_0^M \sum_{n=0}^{\infty} \mathbf{1}_{\{\rho_n \leq M\}} \delta(s - \rho_n) ds \right] = p \cdot R_\tau$$

Define the times between successful packet receptions $\rho_{(p,\Delta)}$

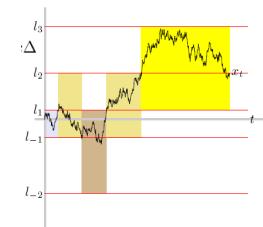
$$\text{Average cost } J_p = \limsup_{T \rightarrow \infty} \frac{1}{T} \mathbb{E} \left[\int_0^T x_s^2 ds \right] = \frac{\mathbb{E} \left[\int_0^{\rho_{(p,\Delta)}} x_s^2 ds \right]}{\mathbb{E} [\rho_{(p,\Delta)}]}$$

Event-based control with losses

Theorem

If packet losses are i.i.d. with probability p ,
then level-triggered sampling gives

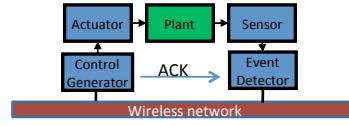
$$J_p = \frac{\Delta^2 (5p + 1)}{6 (1 - p)}$$



Event-based control better than periodic control if loss probability

$$p < 0.25$$

Communication acknowledgements



If controller perfectly acknowledges packets to sensor,
event detector can adjust its sampling strategy

Let

$$\Delta(l) = \sqrt{l+1} \Delta_0$$

where $l \geq 0$ number of samples lost since last successfully transmitted packet

Gives that $\mathbb{E} [\tau_{i+1}^\dagger - \tau_i^\dagger]$ becomes independent of i .

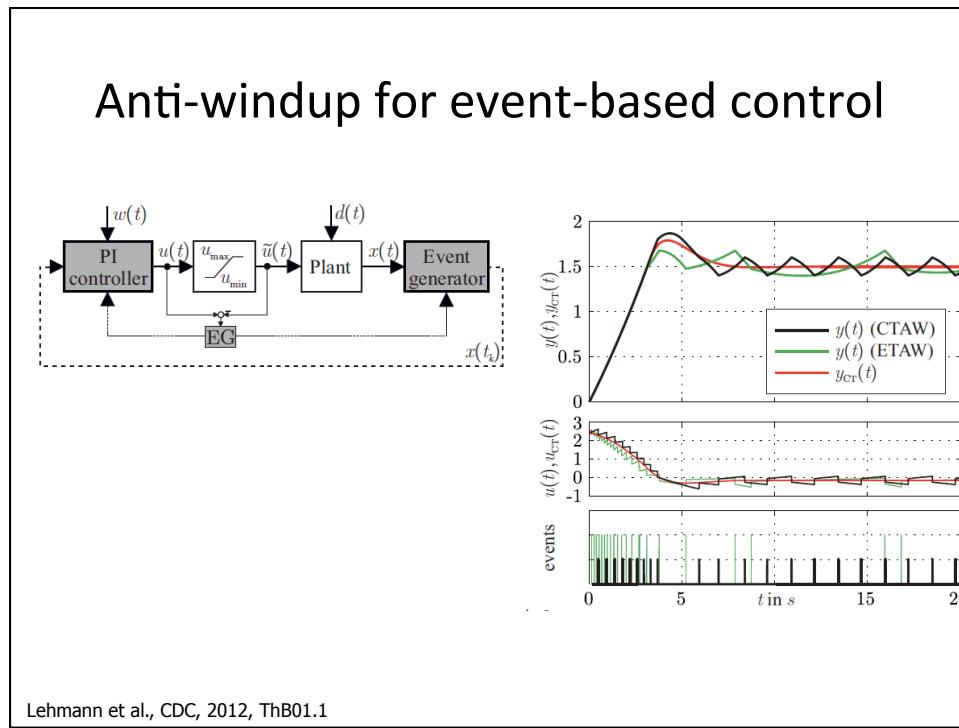
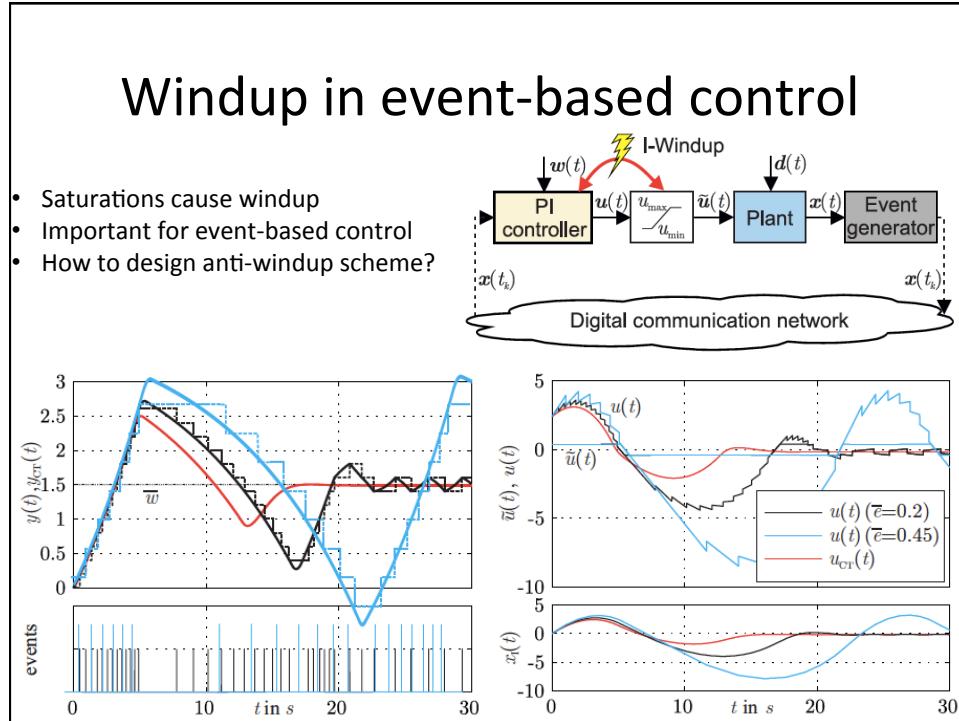
Better performance than fixed $\Delta(l)$ for same sampling rate:

$$J_p^\dagger = \frac{\Delta^2 (1+p)}{6(1-p)} \leq \frac{\Delta^2 (1+5p)}{6(1-p)} = J_p.$$

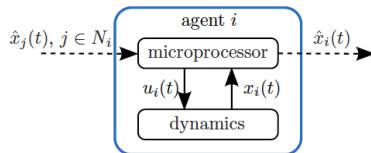
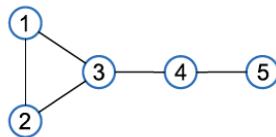
Rabi and J, 2009

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- Extensions
 - Event-based anti-windup
 - Event-based multi-agent systems
- Conclusions



Event-based Control of Multi-Agent System



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

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

Event-based broadcasting

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

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

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

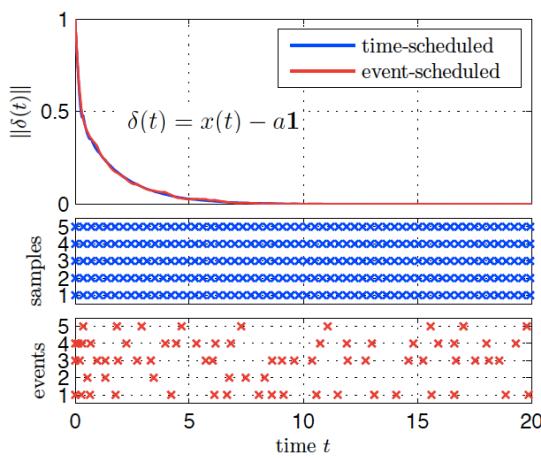
$$f_i(t, e_i(t)) = |e_i(t)| - (c_0 + c_1 e^{-\alpha t})$$

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

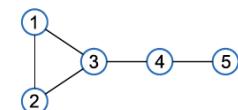
Practical consensus is achieved if $0 < \alpha < \lambda_2(L)$

Seyboth et al. (2011)

Event-based vs Periodic Communication



Graph:



Sampling periods:

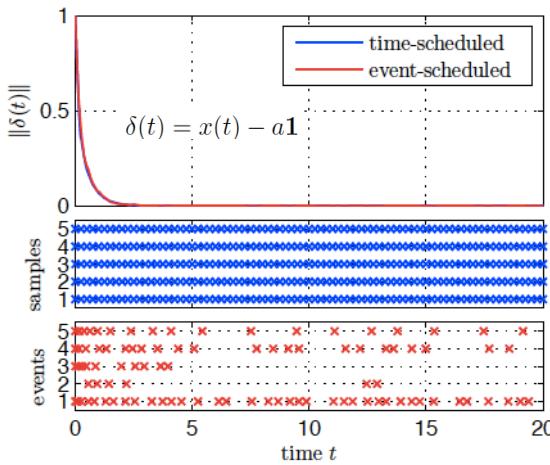
■ Time-scheduling:
 $\tau_s = 0.350$
 $\tau_{max} = 0.480$

■ Event-scheduling:
 $\tau_{mean} = 1.389$

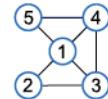
τ_{max} : largest stabilizing sampling period, see G. Xie et al., ACC2009

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)

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Conclusions

- Event-based control is an **enabler for applications** of wireless networked control systems
- Efficient use of **network resources** under control objectives
- **Stochastic control** approach is natural because of probabilistic guarantees for wireless networks
- Many open problems related to **multi-loop** systems and **multi-hop** networks



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