



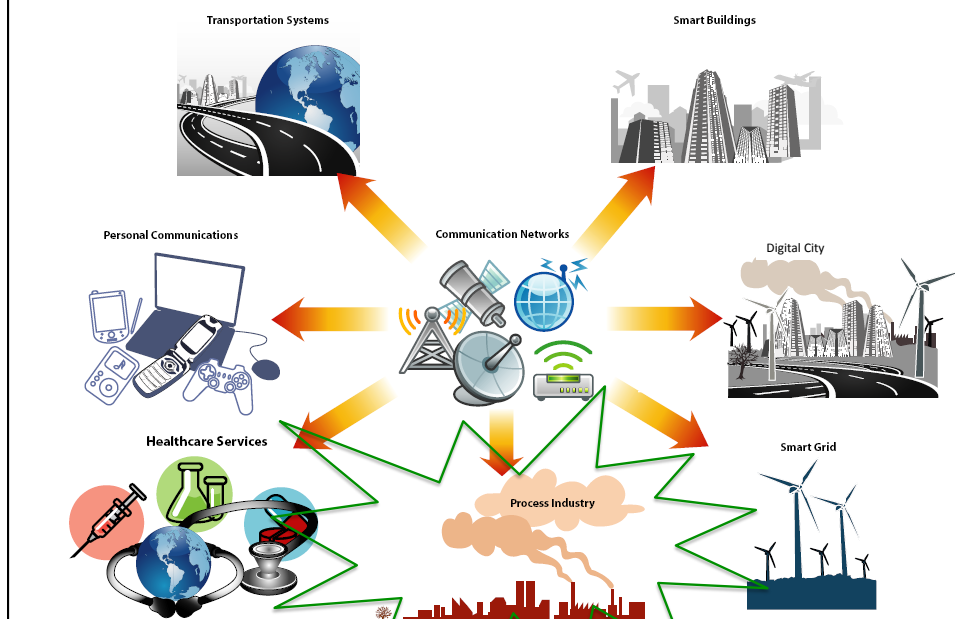
Wireless Control: Medium Access and Event-based Control

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Academy of Mathematics and Systems Science, CAS, Beijing, Jul 4, 2012

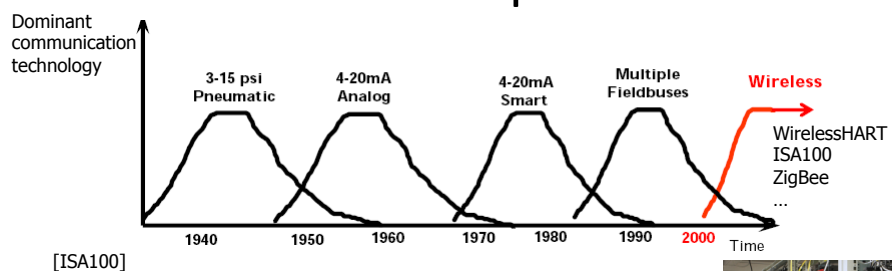
Large-scale networked control systems applications



Outline

- Introduction
- Industrial applications of wireless control
- Medium access
- Event-based control
- Conclusions

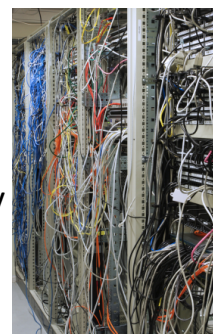
Communication in process control

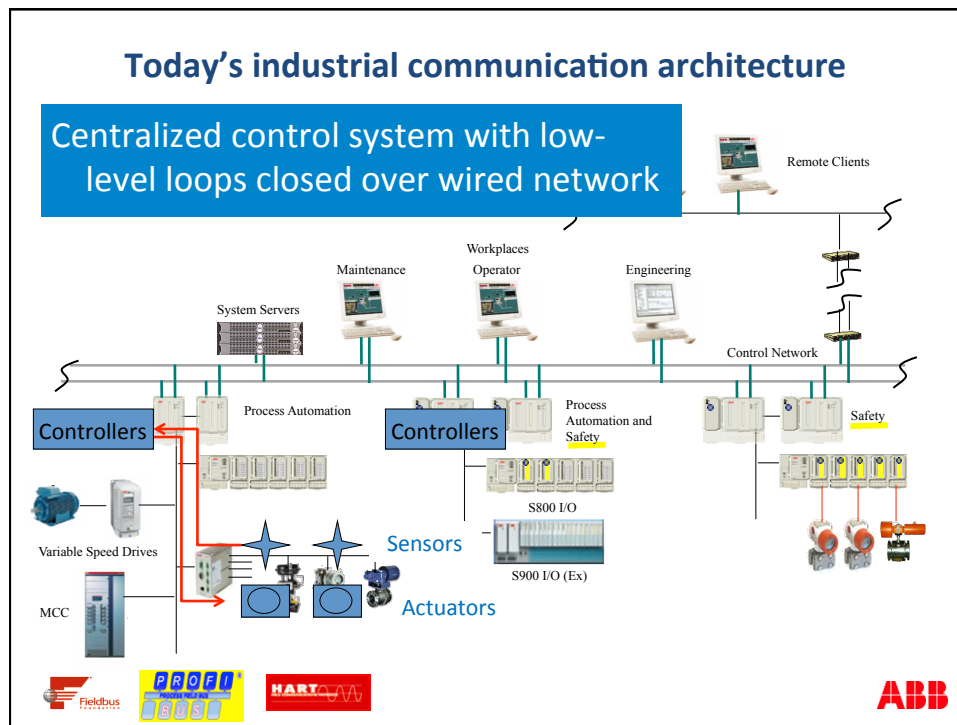


Wireless sensor systems benefit from

- Lower installation and maintenance costs
- Increased sensing capabilities and flexibility

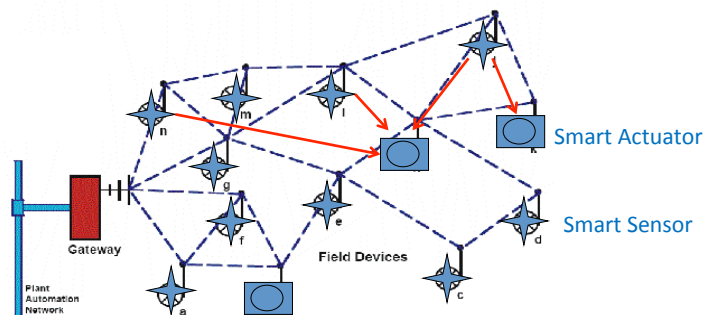
Major consequences for control system architectures





Towards wireless sensor and actuator network architecture

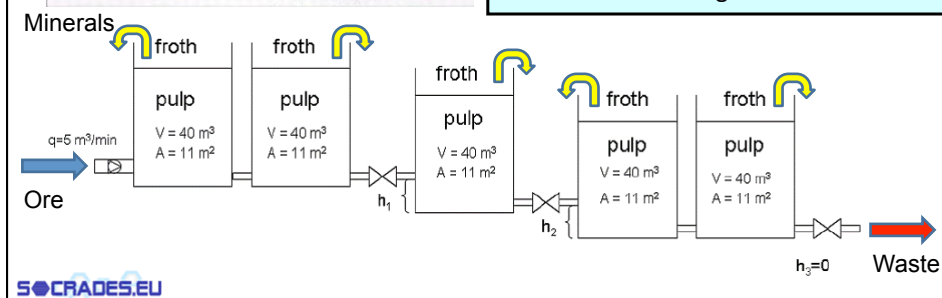
- Local control loops closed over **wireless multi-hop network**
- Potential for a dramatic change:
 - From fixed hierarchical centralized system to flexible distributed
 - Move intelligence from dedicated computers to sensors/actuators



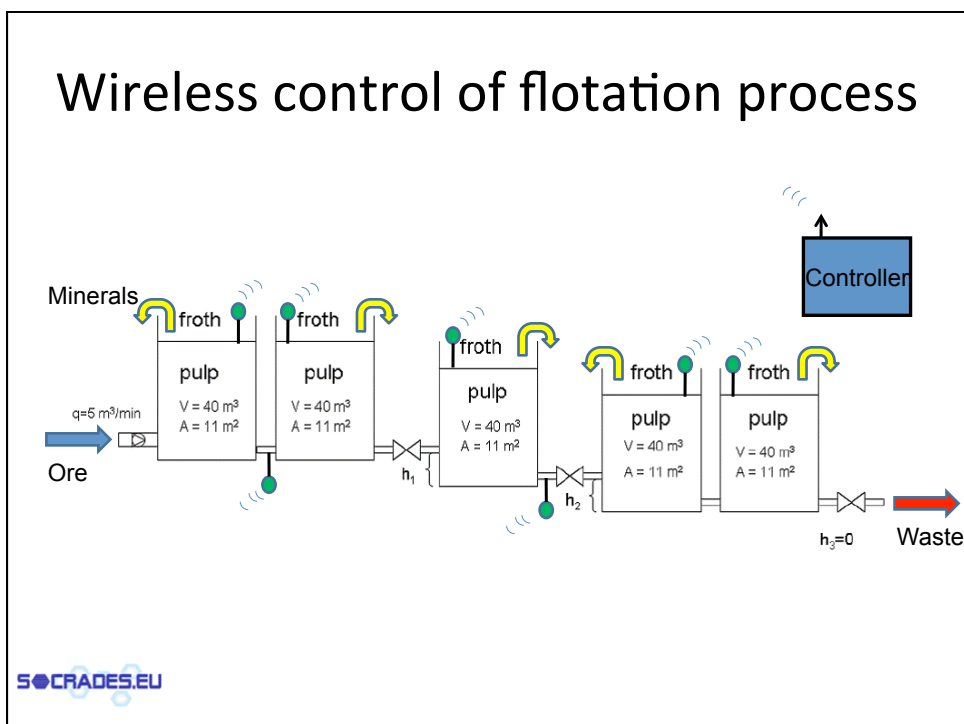
Control of froth flotation process



- Froth flotation process concentrates the metal-bearing mineral in the ore



Wireless control of flotation process



Wireless control of floatation process

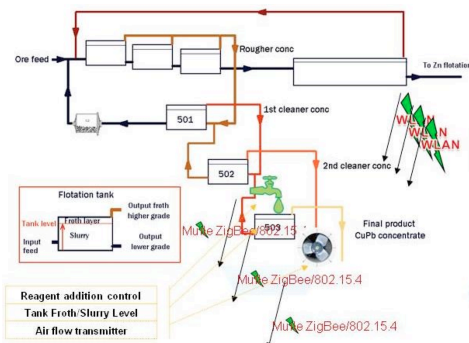
The Boliden plant



Existing wired communication system



Wireless communication for tank level control



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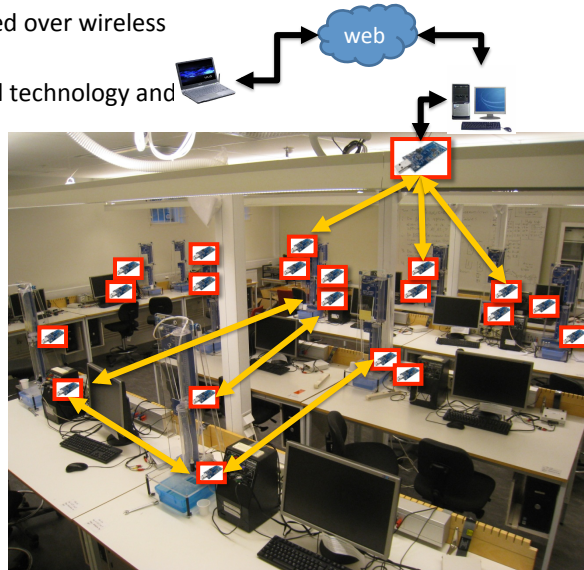
BOLIDEN

ABB

Test-bed for control over IEEE 802.15.4

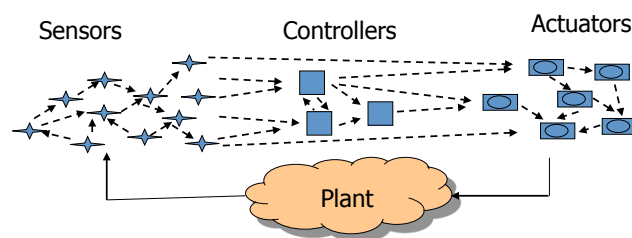
20 coupled water tanks connected over wireless multi-hop network

Test-bed to evaluate new control technology and wireless network protocols



Wireless control system

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



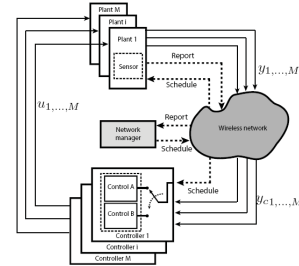
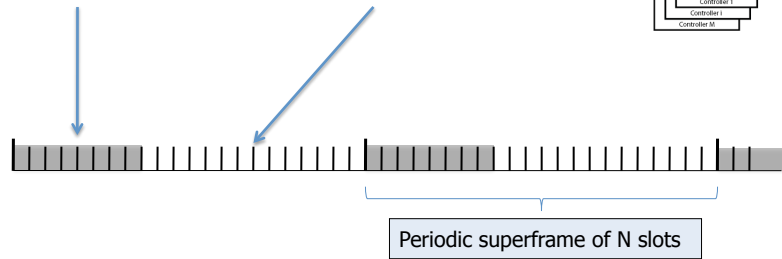
- How handle network imperfections: resource constraints, loss, conflicts, delays, outages?
- How move intelligence from a few central units to many distributed devices?

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Hybrid MAC protocols

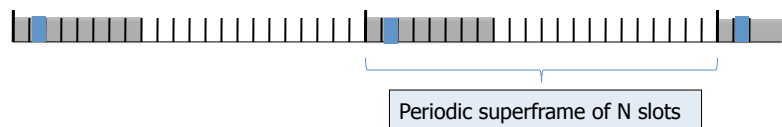
MAC protocols have often both
contention-free and contention access periods



Hybrid MAC protocols

Today's MAC protocol standards have both
 contention-free and contention access periods

Contention-free period for TDMA scheduled communication

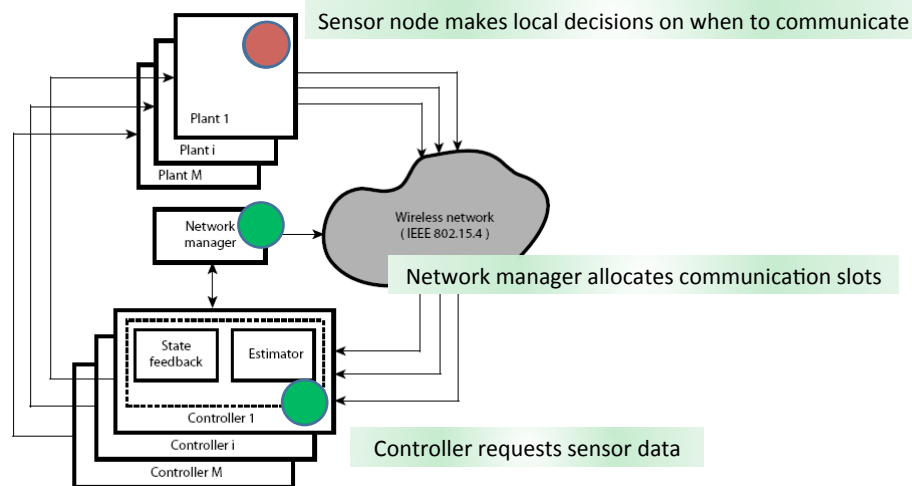


Contention access period for random CSMA communication

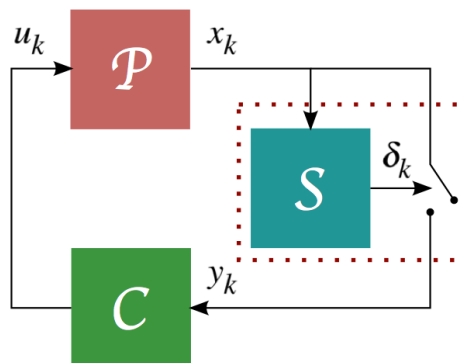


TDMA = Time division multiple access, CSMA/CA = Carrier Sense Multiple Access with Collision Avoidance

Where to take medium access decisions?



Is there a separation principle for scheduling-estimation-control?



Stochastic control formulation

Plant:

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

Scheduler:

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

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

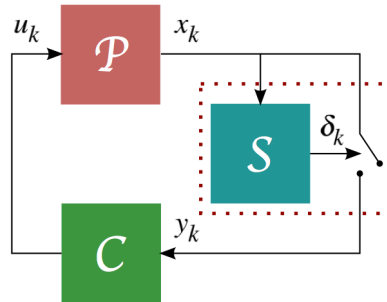
Controller:

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

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

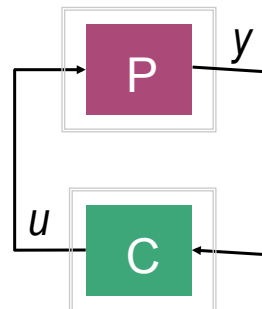
Cost criterion:

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



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 $\hat{x}_{k|k} = E[x_k | \mathbb{I}_k^C]$.



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

State-based scheduler

Plant:

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

Scheduler:

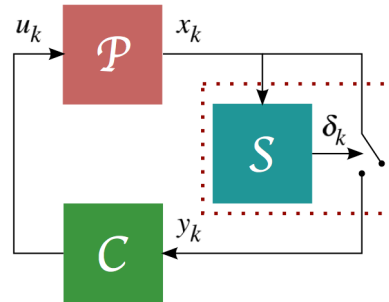
$$\delta_k = f_k(\mathbb{I}_k^S) \in \{0, 1\}$$

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

Controller:

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

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



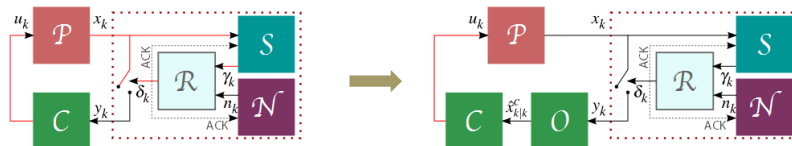
Corollary The control u_k for the optimal closed-loop system has a dual effect.

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

Ramesh, Sandberg, Bao, J, 2009, 2010

Conditions 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 and only if the scheduling decisions are not a function of the applied controls.



Ramesh, Sandberg, Bao, J, 2011

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Observer-Scheduler for Certainty Equivalence

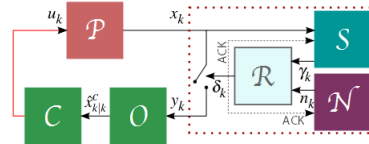
A symmetric scheduling policy results in separation between the estimator and the scheduler, as well as an optimal certainty equivalent

- Observer:
$$\hat{x}_{k|k} = \begin{cases} x_k & \delta_k = 1 \\ A^{k-\tau_k} x_{\tau_k} + \sum_{s=1}^{k-\tau_k} A^{s-1} B u_{k-s} \\ + \mathbb{E}[\sum_{s=1}^{k-\tau_k} A^{s-1} w_{k-s} | \hat{f}_k, \dots, \hat{f}_{\tau_k+1} = 0] & \delta_k = 0 \end{cases}$$

- Symmetric scheduler:

$$\gamma_k = f_{|\cdot|} \left(\sum_{s=1}^{k-\tau_{k-1}} A^{s-1} w_{k-s} \right) \quad \forall k,$$

where, $f_{|\cdot|}(r) = f_{|\cdot|}(-r)$

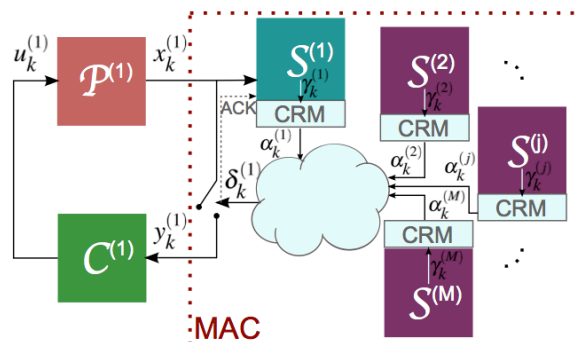


- Certainty equivalence achieved at the cost of optimality

Ramesh, Sandberg, Bao, J, 2011

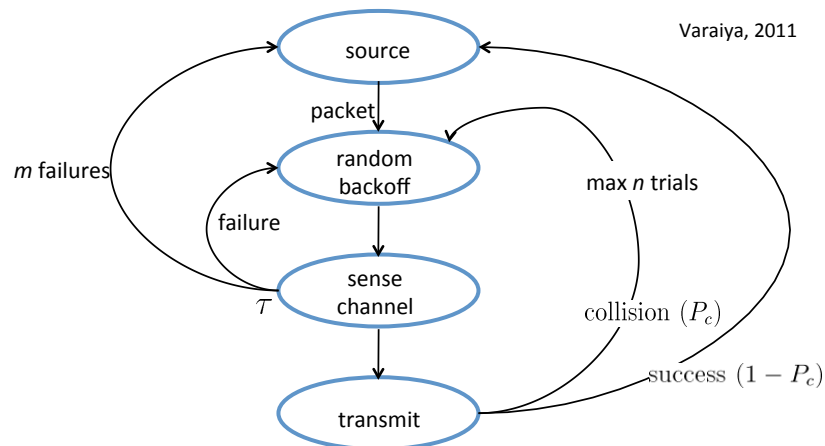
Extensions to multiple loops and contention resolution mechanisms

- Hard problem because of the correlation between the plants imposed by the MAC
- Closed-loop analysis can still be done for a class of event-based schedulers and simple MAC's
- General problem with event-based schedulers and realistic MAC (e.g., CSMA/CA) is open



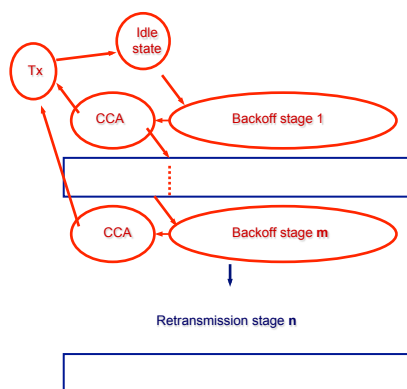
Ramesh et al, 2011

How to model CSMA/CA MAC?



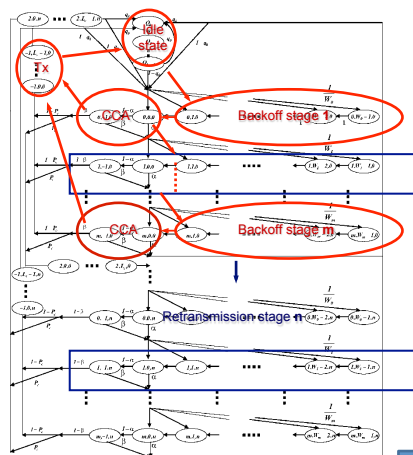
- Every device executes this protocol
- Assume all carrier sense events are independent [Bianchi, 2000]

CSMA/CA mechanism of a node in an IEEE 802.15.4 wireless network



- A transmitting node delays for a random number of backoff periods in $[0, 2^{m_b} - 1]$, where m_b is the **initial backoff exponent**.
- If two consecutive clear channel assessments (CCA) are idle, the node starts the transmission and waits for an ACK
- If the channel is busy, the procedure is repeated increasing the backoff windows until a **maximum backoff exponent m_b** .
- After a **maximum number of backoffs m** the packet is discarded.
- In case of collision the procedure is restarted and repeated until a **retry limit n**

Markov chain model of CSMA/CA



- 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_0, 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

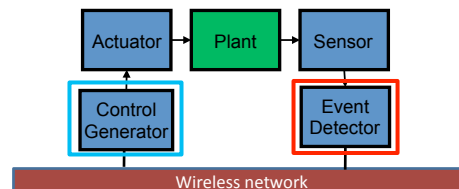
Park, Di Marco, Soldati, Fischione, J, 2009

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When to transmit?

- Medium access control-like mechanism at sensor
 - E.g., threshold crossing, adaptive sampling



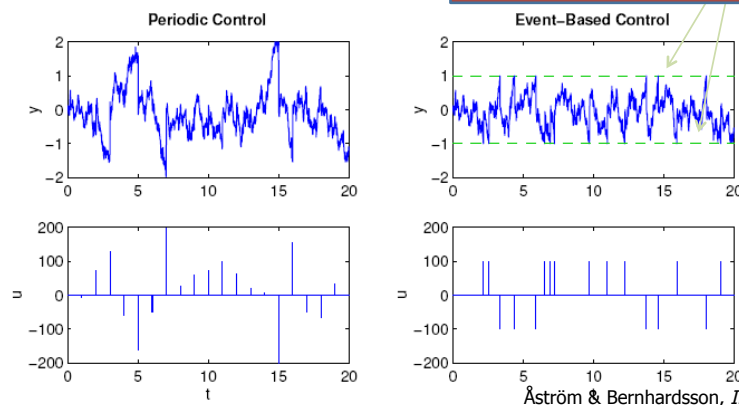
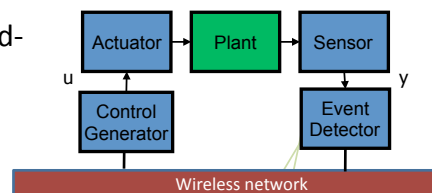
How to control?

- Execute control law over fixed control alphabet
 - E.g., piecewise constant controls, impulse control

Åström, 2007, Rabi and J., WICON, 2008

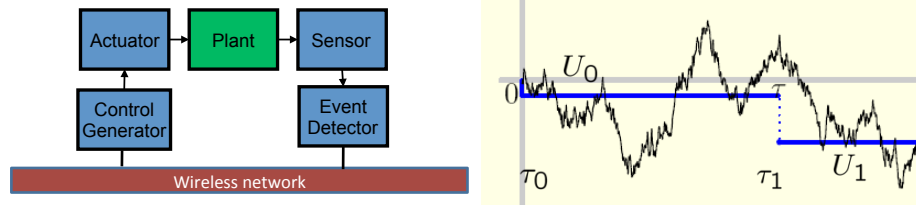
Example: Fixed threshold with impulse control

- Event-detector implemented as fixed-level threshold at sensor
- Event-based impulse control better than periodic impulse control



Åström & Bernhardsson, IFAC, 1999

Event-based ZoH control with adaptive sampling



How choose $\{U_i\}$ and $\{\tau_i\}$ to minimize $V = \frac{1}{T} E \int_0^T x^2(t) dt$.

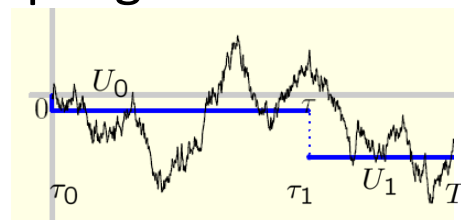
Rabi, J, Johansson, 2008

Controlled Brownian motion with one sampling event

$$dx_t = u_t dt + dB_t$$

$$\min_{U_0, U_1, \tau} J = \min_{U_0, U_1, \tau} \mathbf{E} \int_0^T x_s^2 ds$$

$$= \min_{U_0, U_1, \tau} \left[\mathbf{E} \int_0^\tau x_s^2 ds + \mathbf{E} \int_\tau^T x_s^2 ds \right]$$



A joint optimal control and optimal stopping problem

Rabi, J, Johansson, 2008

$$dx_t = u_t dt + dB_t$$

$$\min_{U_0, U_1, \tau} J = \min_{U_0, U_1, \tau} \mathbf{E} \int_0^T x_s^2 ds$$

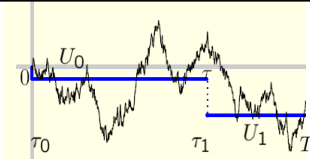
If τ chosen deterministically (not depending on x_t) and $x_0 = 0$:

$$U_0^* = 0 \quad U_1^* = -\frac{3x_{T/2}}{T} \quad \tau^* = T/2$$

If τ is event-driven (depending on x_t) and $x_0 = 0$:

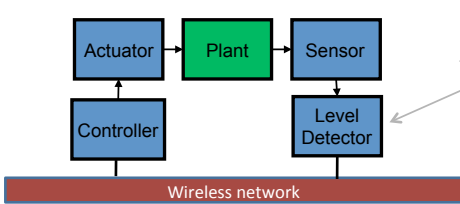
$$U_0^* = 0 \quad U_1^* = -\frac{3x_{\tau^*}}{2(T - \tau^*)}$$


$$\tau^* = \inf\{t : \underbrace{x_t^2}_{\geq \sqrt{3}(T-t)}\}$$



Envelope defines optimal level detector

Optimal level detector





$$dx_t = u_t dt + dB_t$$

$$\min_{U_0, U_1, \tau} J = \min_{U_0, U_1, \tau} \mathbf{E} \int_0^T x_s^2 ds$$

Policy iteration

For $x_0 \neq 0$ we have in general the cost function

$$J_N(x_0, \{U_0, U_1\}, \tau) \triangleq \alpha(x_0, T) - \mathbb{E}[\beta(x_0, U_0, \tau, T)],$$

where

$$\alpha(x_0, U_0, T) = \int_0^T \mathbb{E}[\Phi_{U_0}^2(s, 0, x_0)] ds$$

$$\beta(x_0, U_0, \tau, T) = \int_\tau^T \mathbb{E}[\Phi_{U_0}^2(s, \tau, x_\tau) - \Phi_{U_1^*}^2(s, \tau, x_\tau)]$$

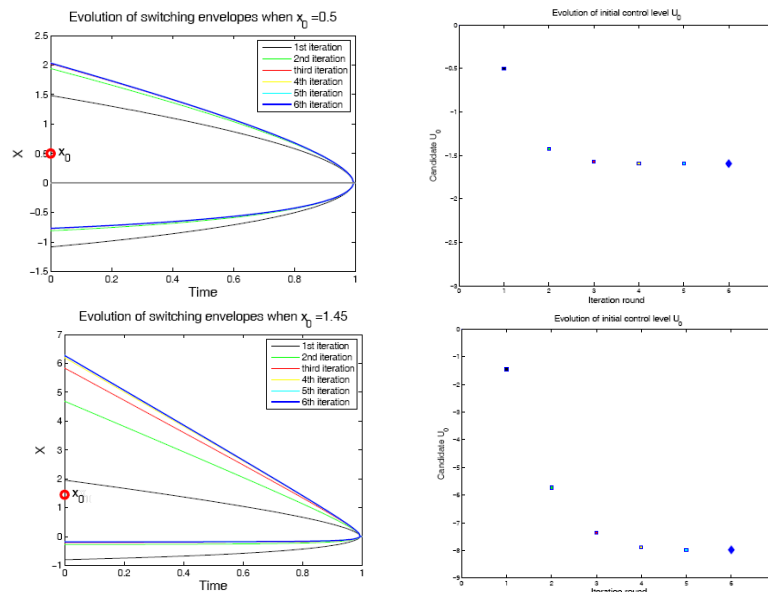
and $\Phi_U(t_2, t_1, x)$ is the solution of the system with constant control

Necessary condition for optimality

$$\begin{cases} \tau^*(x_0) = \operatorname{ess\,sup}_{\tau} \mathbb{E}[\beta(x_0, U_0^*(x_0), \tau, T)], \\ U_0^*(x_0) = \inf_U \left\{ \alpha(x_0, U, T) - \mathbb{E}[\beta(x_0, U, \tau^*(x_0), T)] \right\}. \end{cases}$$

suggests iterative search algorithm. Computationally intensive.

Example: Non-zero initial conditions



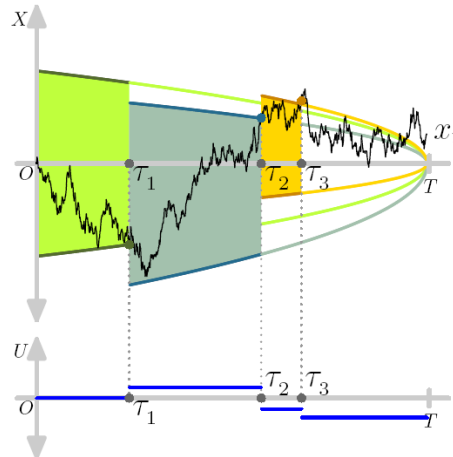
Multiple samples

Extension to $N > 1$ samples

$$J_N(x_0, \mathcal{U}, \{\tau\}_{i=1}^N) = \mathbb{E} \left[\int_0^T x_s^2 ds \middle| x_0 \right]$$

through nested single sample problems

Extensions to variable budget sampling, allowing number of samples to depend on x .



Event-based impulse control

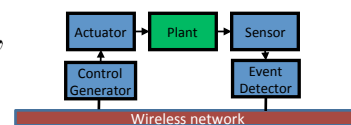
Plant $dx_t = dW_t + u_t dt, x(0) = x_0,$

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

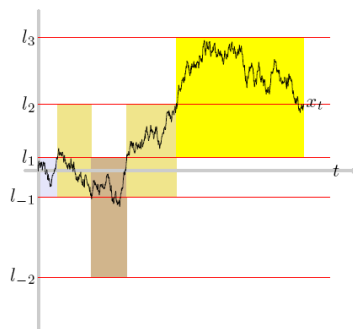


Level-triggered control

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



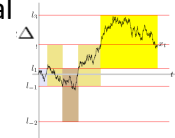
Level-triggered control

For Brownian motion, equidistant sampling is optimal

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

First exit time

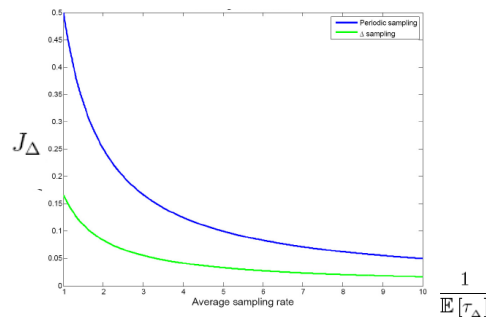
$$\tau_\Delta = \inf \{ \tau \geq 0, x_\tau \notin (\xi - \Delta, \xi + \Delta), x_0 = \xi \}$$



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, 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 $\rho_i \in \{\tau_0 = 0, \tau_1, \tau_2, \dots\}$,

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

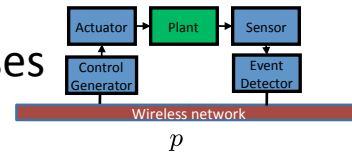
Average rate of packet reception

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

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

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

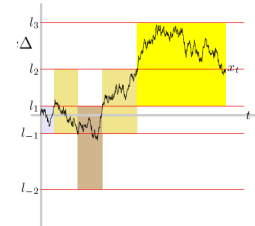
Event-based control with losses



Theorem

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

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

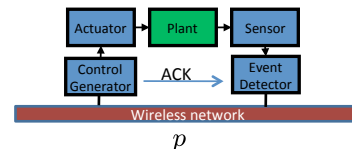


Event-based control better than periodic control if loss probability

$$p < 0.25$$

Rabi and J, 2009

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

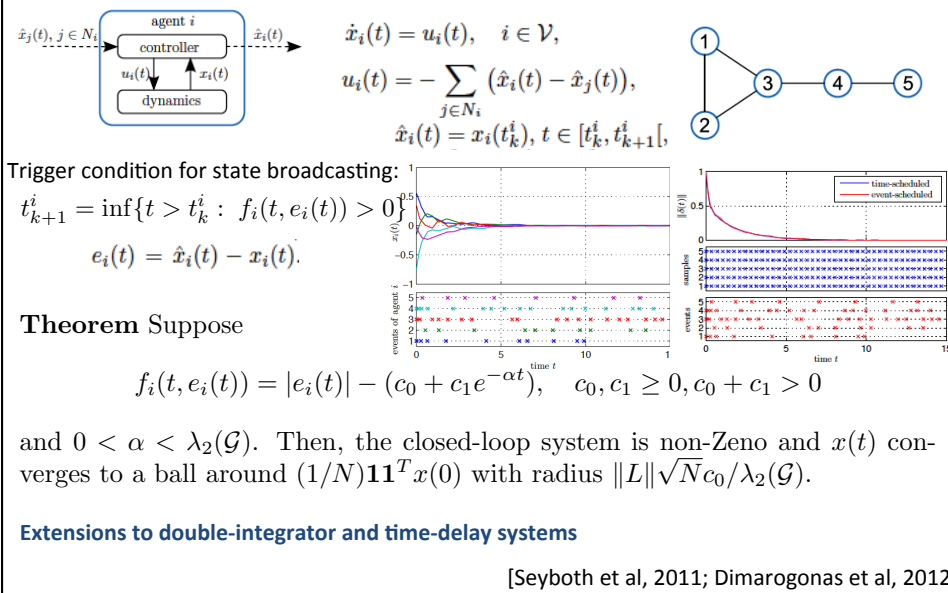
Gives that $\mathbb{E}[\tau_{i+1}^\dagger - \tau_i^\dagger]$ becomes independent of i .

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

Event-based Control for Multi-agent Systems



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Conclusions

- **Wireless control** is an enabling technology in many emerging industrial applications
- Fundamental challenges related to
 - **time-driven**, synchronous, sampled data control theory, vs
 - **event-driven**, asynchronous, ad hoc wireless networking
- Integrated modeling for medium access and control
- Event-based control provides a natural principle for large-scale wireless control systems

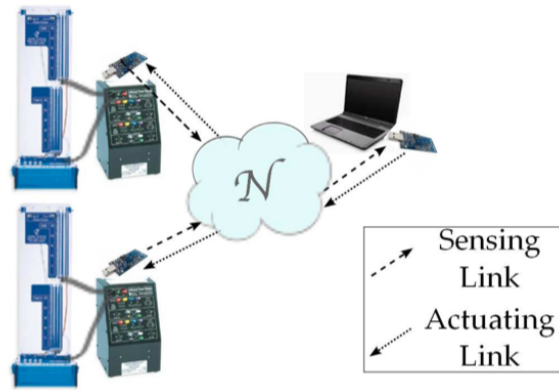
<http://www.ee.kth.se/~kallej>

Control over wireless network

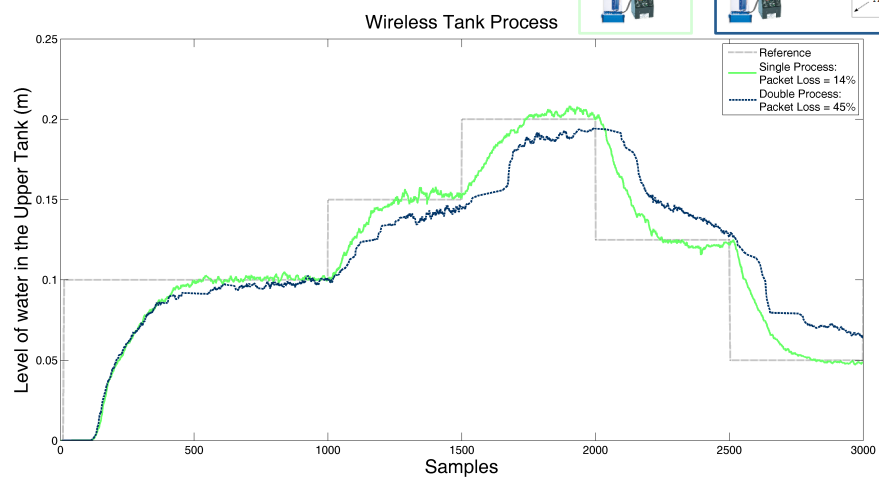
Single Process



Double Process

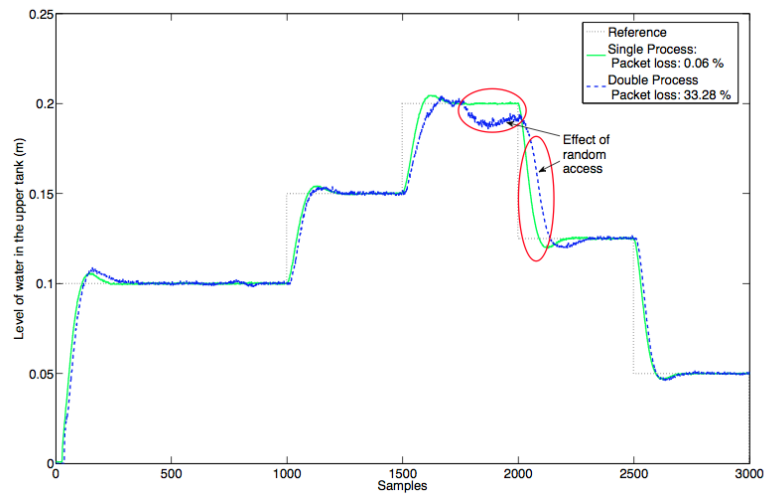
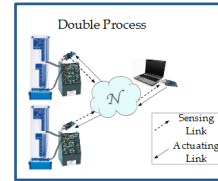


Packet loss influence on control performance



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Partial improvement using CSMA/CA medium access



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