



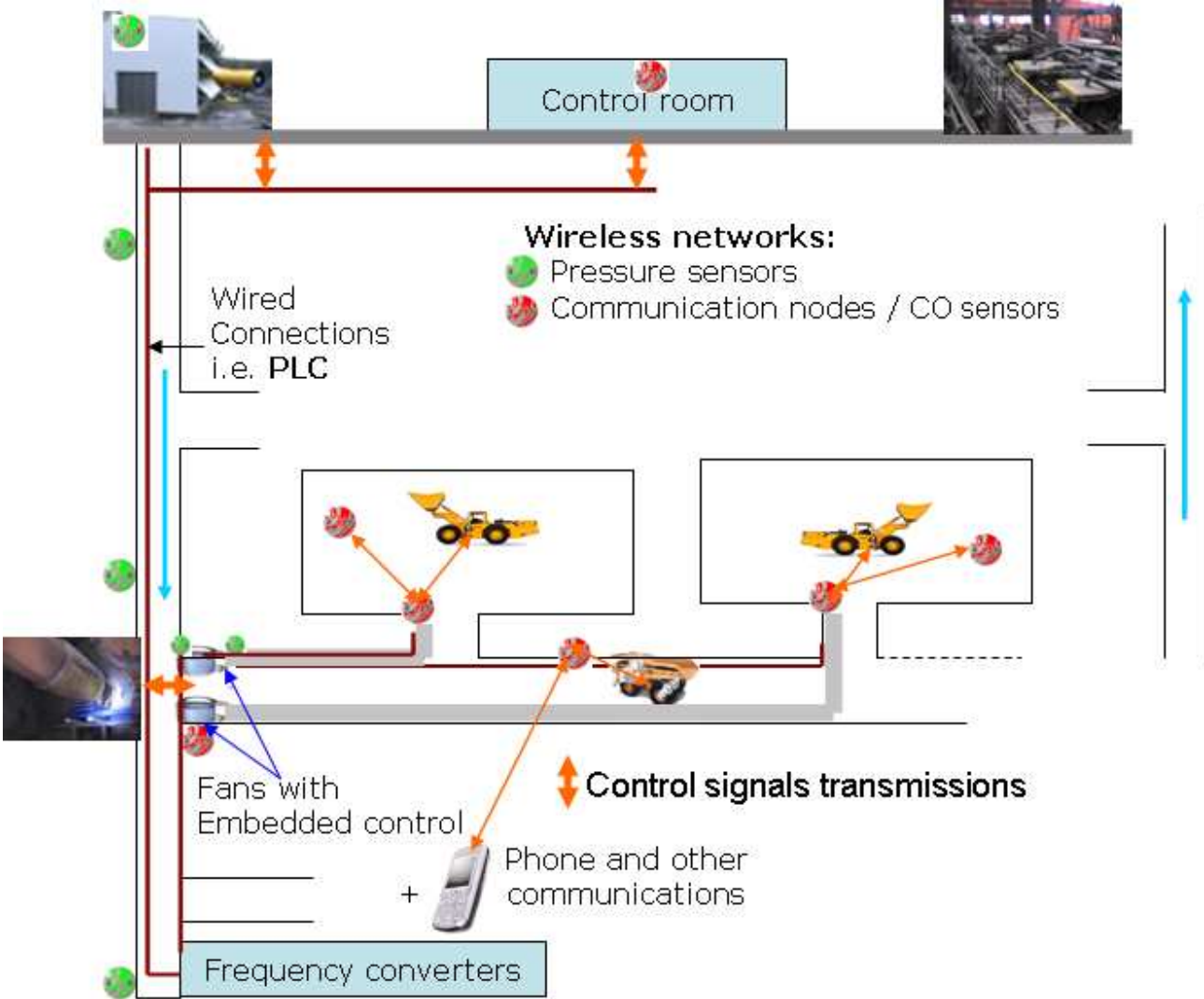
KTH Electrical Engineering

Predictive control over wireless multi-hop networks

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Illustrative example: mining ventilation problem (HYCON)



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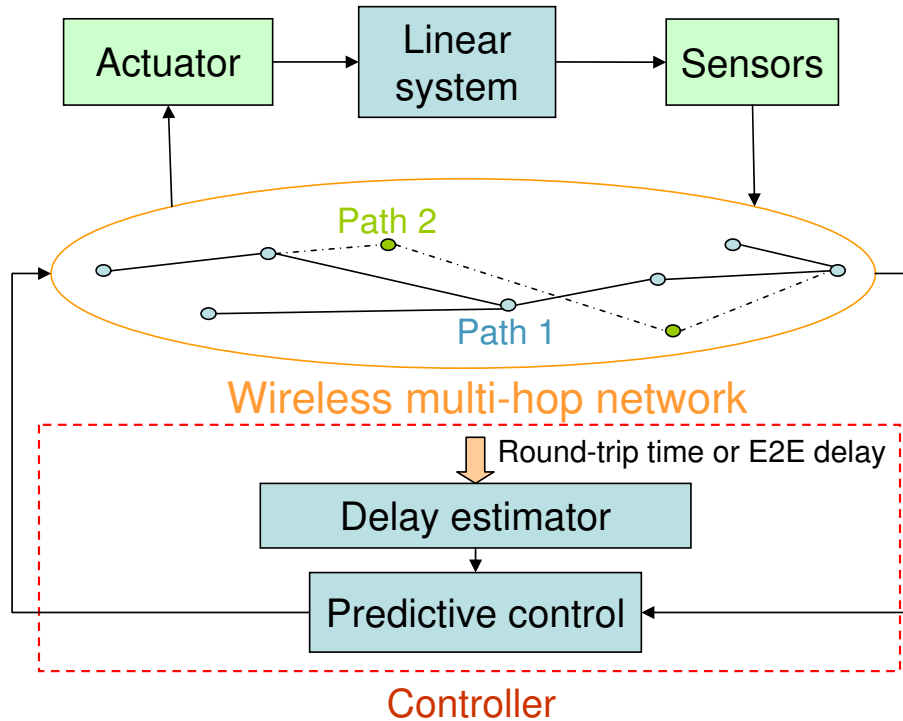


University of L'Aquila

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I. Problem description



- Time delays:
 - $\tau_1(t)$: sensors \rightarrow controller;
 - $\tau_2(t)$: controller \rightarrow actuator.
- Packet losses:

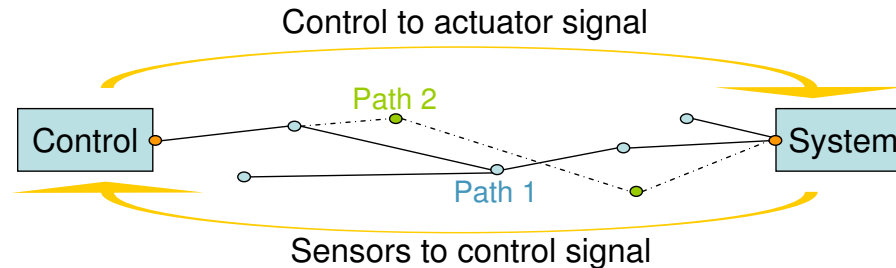
$$\begin{cases} w(t_k) = u(t_k - \tau_2(t_k)) & \text{if received,} \\ w(t_k) = w(t_{k-1}) & \text{else.} \\ v(t) = w(t_k) & \forall t \in [t_k, t_{k+1}) \end{cases}$$
- Bandwidth limitation.

- Linear system:

$$\dot{x}(t) = Ax(t) + Bv(t), \quad x(0) = x_0$$

$$y(t) = Cx(t)$$

II. Wireless multi-hop networks



- Network characteristics:
 - time-varying channel and network topology;
 - dynamic selection of $h(t)$ hops;
 - next node ensures progress toward destination;
 - i.e. random sleep to save energy.

- Delay associated with each node i :

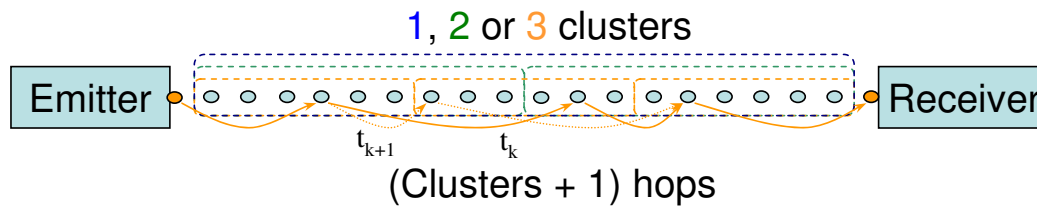
α_i : time to wait before sending a packet (random);

F : propagation and transmission;

β_i : Automatic Repeat reQuest (ARQ): retransmission;

$$\Rightarrow \tau(t) = h(t)F + \sum_{i=1}^{h(t)} (\alpha_i + \beta_i)$$

i.e. clusters and dynamic selection of hops



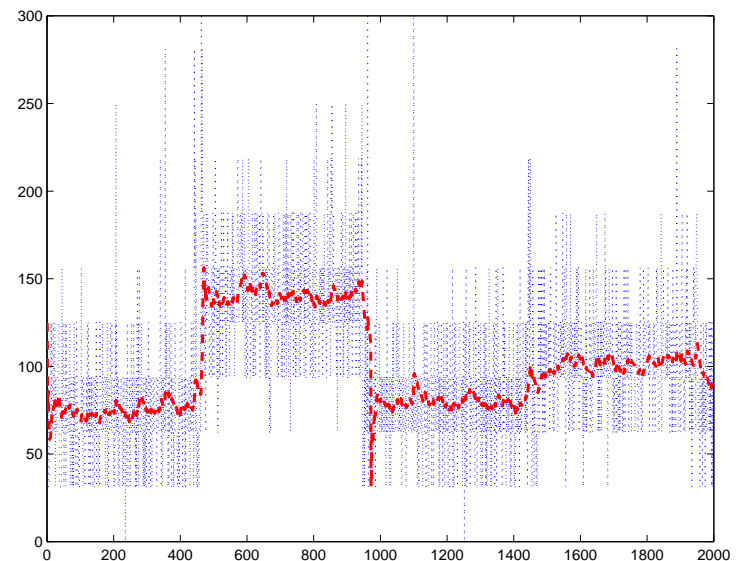
- sleeping strategy depending on traffic and network conditions,
- optimize energy consumption in a clustered environment [Bonivento & al.'06],

⇒ Dynamical organization in clusters [P.G. Park MS'07].

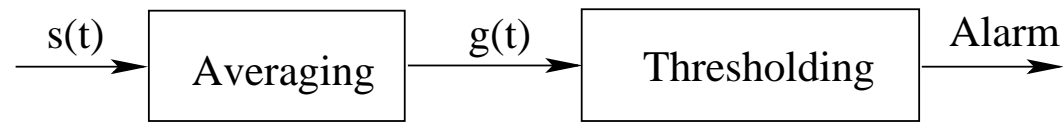
Three possible behaviors:

- Low freq. = changes in the number of clusters,

- Medium freq. = node selection within a cluster,
- High freq. = transmissions between nodes.



III. Time-delay estimator.



Kalman filter + CUMSUM change-detection [Gustafsson'00, Jacobsson & al.'04]:

- measure RTT or 1-channel delay = large variability and persistent changes in the mean,
- smooth out fast variations and react quickly to changes in mean,

⇒ sets the time-varying horizon of the predictive controller.

The signal model is

$$\begin{aligned} \tau_{k+1} &= \tau_k + v_k \\ r_k &= c\tau_k + w_k \end{aligned} \quad , \quad v_k = \begin{cases} 0 & \text{with probability } 1 - q, \\ \nu & \text{with probability } q, \text{ where } \text{Cov}(v_k) = q^{-1}Q_k \end{cases}$$

Kalman filter updates

$$\begin{aligned}\hat{\tau}_{k+1} &= \hat{\tau}_k + K_k(r_k - c\hat{\tau}_k) \\ K_k &= \frac{cP_{k-1}}{R_k + c^2P_{k-1}} \\ P_k &= P_{k-1} - \frac{c^2P_{k-1}^2}{R_k + c^2P_{k-1}} + Q_k\end{aligned}$$

CUMSUM detector

$$\begin{aligned}\varepsilon_k &= r_k - c\hat{\tau}_k \\ g_k^+ &= \max(g_{k-1} + \varepsilon_k - \kappa, 0) \\ g_k^- &= \max(g_{k-1} - \varepsilon_k - \kappa, 0)\end{aligned}$$

if g_k^+ or $g_k^- >$ threshold h , P_k set to a large value, g^+ and g^- reset to zero, and $\hat{\tau}_k$ set to $c^{-1}r_k$.

IV. Predictive control approach.

Ideal case: known delay [Witrant & al.'05...]

$$\dot{x}(t) = Ax(t) + Bu(t - \tau_2), \quad x(0) = x_0$$

with (A, B) a controllable pair, $\tau_{1,2} \in C(\mathbb{R}^+, [0, \tau_{max}])$, $\dot{\tau}_{1,2} < 1$, $\forall t \geq t_0$.
Then the feedback control law

$$u(t) = -Ke^{A\delta} \left[e^{A\tau_1} x(t - \tau_1) + e^{At} \int_{t-\tau_1}^{t+\delta} e^{-A\theta} Bu(\theta - \tau_2(\theta)) d\theta \right]$$
$$\dot{\delta}(t) = -\frac{\lambda}{1 - d\tau_2(\zeta)/d\zeta} \delta + \frac{d\tau_2(\zeta)/d\zeta + \lambda\tau_2(\zeta)}{1 - d\tau_2(\zeta)/d\zeta}$$

with $\zeta = t + \delta$, $\lambda > 0$, $\delta(0) = \delta_0$, ensures that the system trajectories $x(t)$ converge exponentially to zero.

⇒ *time-shifted* dynamics

$$\frac{dx}{d\zeta} = (A - BK)x(\zeta)$$

and *equivalent system* [Artstein'82] spectrum

$$A - \left| \frac{1 + \dot{\delta}}{\dot{\tau}(t + \delta)} \right| e^{-A\delta} BK,$$

Worst case (for design):

$$A - \left| \frac{1}{1 - \dot{\tau}_{max}} \right| e^{-A\tau_{max}} BK$$

is Hurwitz.

Use of the delay estimator

i.e. transmission over the same network: $\hat{\tau} \approx \hat{\tau}_1 \approx \hat{\tau}_2$

$$u(t) = -Ke^{A\delta} \left[e^{A\hat{\tau}} x(t - \tau_1) + e^{At} \int_{t-\hat{\tau}}^{t+\hat{\delta}} e^{-A\theta} Bu(\theta - \hat{\tau}(\theta)) d\theta \right]$$

$$\dot{\hat{\delta}}(t) = \frac{\dot{\hat{\tau}}(t) + \lambda(\hat{\tau}(t) - \hat{\delta}(t))}{1 - \dot{\hat{\tau}}(t)}$$

and causal integral approximation:

$$\mathcal{I}_k = e^{At_k} \int_{t_k - \hat{\tau}_k}^{t_k + \hat{\delta}_k} e^{-A\theta} Bu(\theta - \hat{\tau}(\theta)) d\theta = e^{A\hat{\tau}_k} \int_0^{\hat{\tau}_k + \hat{\delta}_k} f(\mu) d\mu, \quad \text{with}$$

$$f(\mu) = e^{-A\mu} Bu(\min(\mu + t_k - \hat{\tau}_k - \hat{\tau}(\min(\mu + t_k - \hat{\tau}_k, t_k)), t_k))$$

⇒ High frequency noise on control signal when real wireless network data.

Discretization (safe implementation [Mondié & al.'03]):

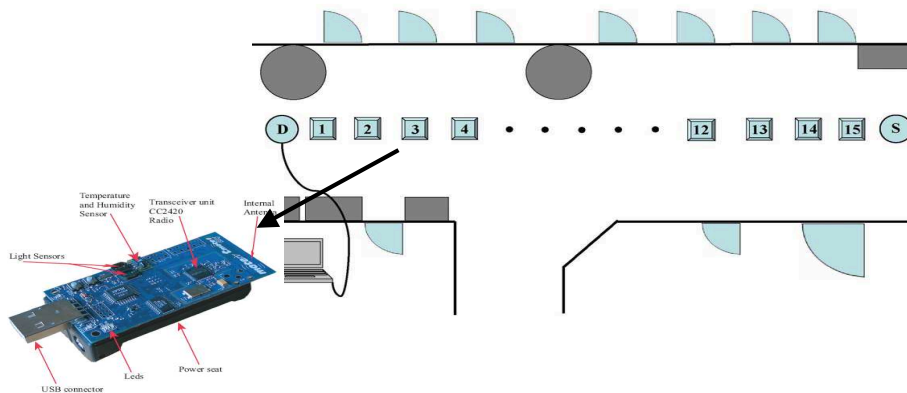
$$\begin{cases} \dot{z}(t) &= -az(t) + bKe^{A(\hat{\delta}_k + \hat{\tau}_k)} \left[x(t - \tau_1) + \frac{t_s}{2} \sum_{i=1}^{n_k} f((i-1)t_s) + f(it_s) \right] \\ u_k &= z(t_k), \quad n_k \doteq \frac{\hat{\delta}_k + \hat{\tau}_k}{t_s} \end{cases}$$

and filter the triggers with the horizon dynamics

$$\dot{\hat{\delta}}(t) = \frac{\dot{\hat{\tau}}_k + \lambda(\hat{\tau}_k - \hat{\delta}(t))}{1 - \dot{\hat{\tau}}_k}, \quad \text{where} \quad \begin{cases} \dot{\hat{\tau}}_k &= \frac{\hat{\tau}_k - \hat{\tau}_{k-1}}{t_s} & \text{if } \dot{\hat{\tau}} < \dot{\hat{\tau}}_{thr} \\ \dot{\hat{\tau}}_k &= \dot{\hat{\tau}}_{thr} & \text{else} \end{cases}$$

V. Experimental and simulation results

Network setup and experimental measurements

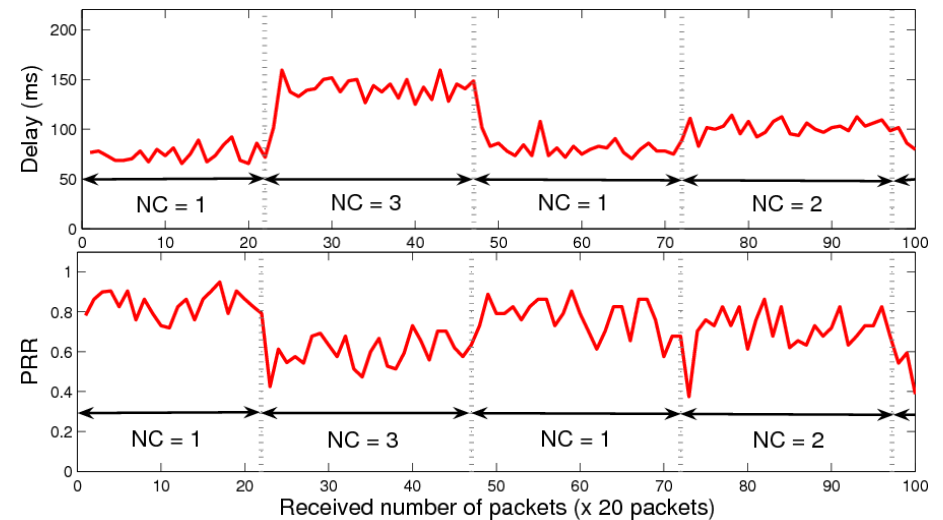


– 20 packets/s

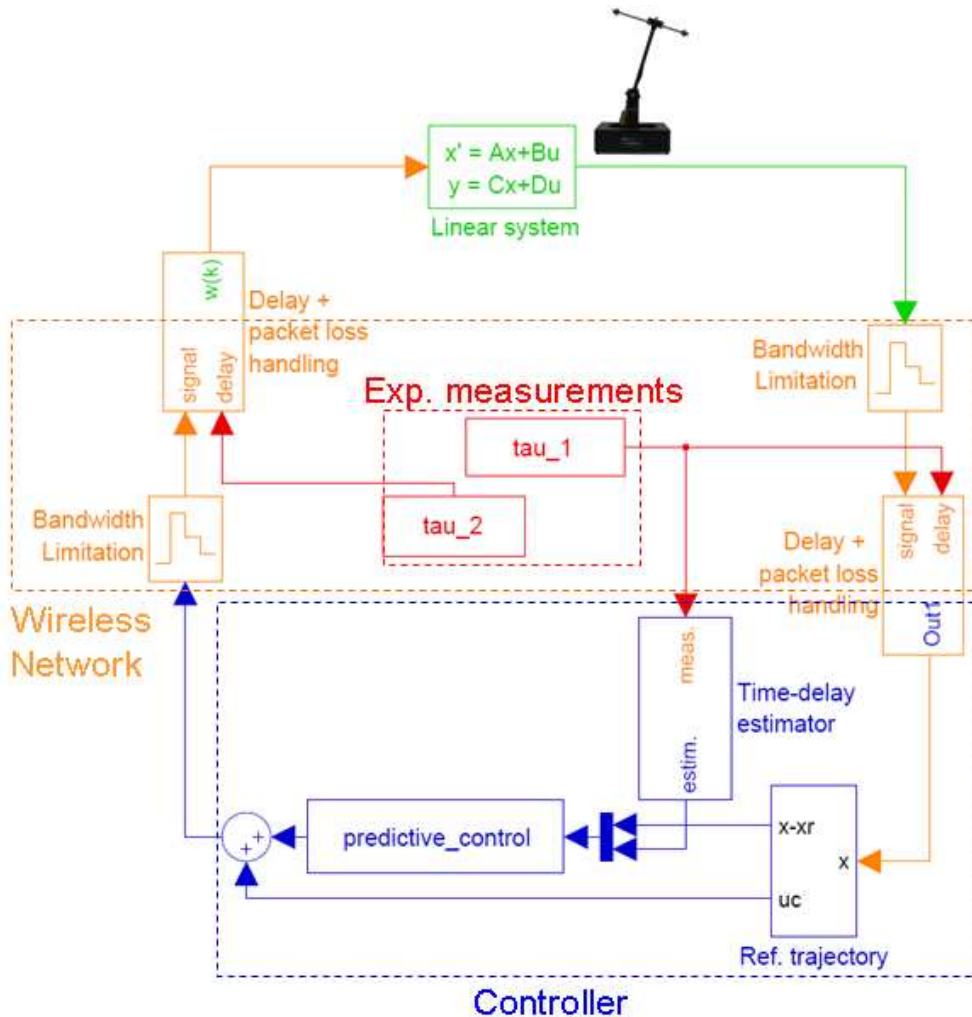
- Randomized Protocol

- Tmotes Sky nodes

- radio controller Chipcon
- CC2420 (2.4GHz)
- IEEE 802.15.4



Control setup



Linear system:

$$\dot{x}(t) = A x(t) + B u(t)$$

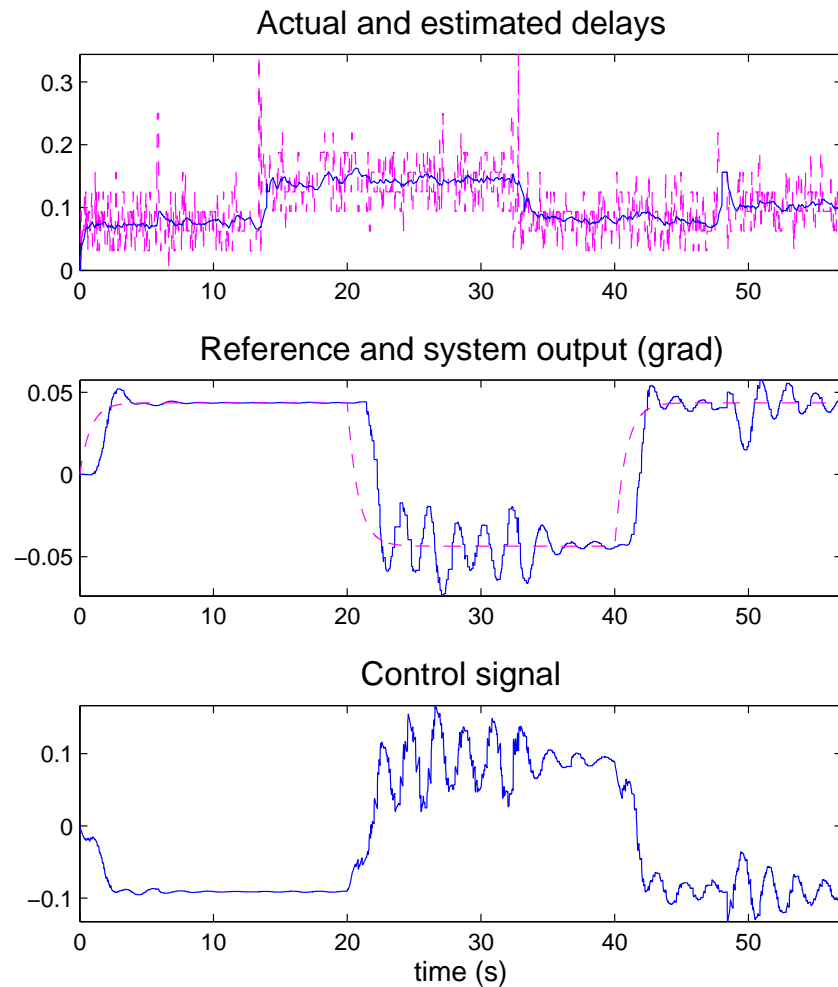
$$y(t) = x(t)$$

with

$$A = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0.80 & 0 & 12.56 & 0 \\ 0 & 0 & 0 & 1 \\ -2.42 & 0 & -8.33 & 0 \end{bmatrix}, \quad B = \begin{bmatrix} 0 \\ 4.57 \\ 0 \\ 0.38 \end{bmatrix}$$

$\lambda_{1,2} = 0.74 \pm 2.08i$, $\lambda_{3,4} = -0.74 \pm 2.08i$, K sets the poles to $(-3 + 0.1i, -3 - 0.1i, -3.5, -6)$.

Simulation results



- Proportional and *non-safe* predictor do not stabilize this system (divergence when 3 clusters)
- Proper delay estimation and compensation
- Stabilization even with 30% of lost packets, bounded oscillation when 40%
- Still one degree of freedom with K

Conclusions

- Control strategy:
 - multi-hop wireless network = time-varying delays, packet losses and bandwidth limitations,
 - delay from CUMSUM Kalman filter: fast mean tracking + HF rejection,
 - ⇒ causal dynamic predictive control compensating the estimated delay.
- Simulation results:
 - experimental delay and packet loss data,
 - efficiency to control over noisy communication channels,
 - limitations: effect of packet losses and protocol.
- Perspectives:
 - wireless multi-hop network design for large-scale applications,
 - optimal design to set K depending on packet losses.

Thanks!

Minimized cost function:[P.G. Park'07]

$$\begin{aligned}
 & \text{Arg min}_h T \left(\lambda \left(\underbrace{h E_{delay}}_{\text{Rand. contension}} + \underbrace{h E_{Re} l_m}_{\text{reception}} + \underbrace{h E_{Te} l_m + \epsilon_m l_m D^\beta h^{1-\beta}}_{\text{send data}} \right) \right. \\
 & \qquad \qquad \qquad \underbrace{\hspace{15em}}_{\text{transmission/reception}} \\
 & \qquad \qquad \qquad + \underbrace{\frac{\max \{D_c(h), E_c(h)\}}{p}}_{\text{wake up}} \left(\underbrace{h E_{ac}}_{\text{active time}} + \underbrace{\epsilon_b l_b D^\beta h^{1-\beta}}_{\text{send beacon}} \right) \\
 & \qquad \qquad \qquad \underbrace{\hspace{15em}}_{\text{Wake up \& beaconing}}
 \end{aligned}$$