

About the stability of sampled-data systems with non-uniform sampling

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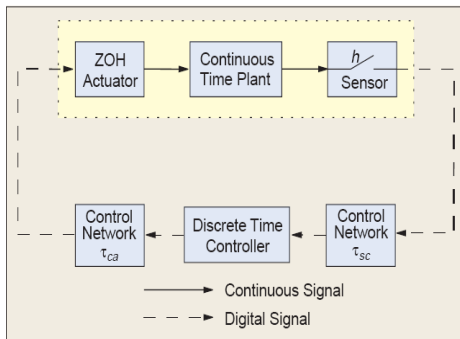
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Motivating problem : Digital control

Classical control loop

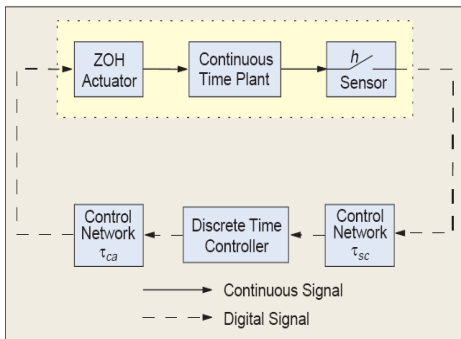


Ideal Hypothesis :

- ▶ Sampling and actuation are periodic and synchronous

Motivating problem : Digital control

Classical control loop

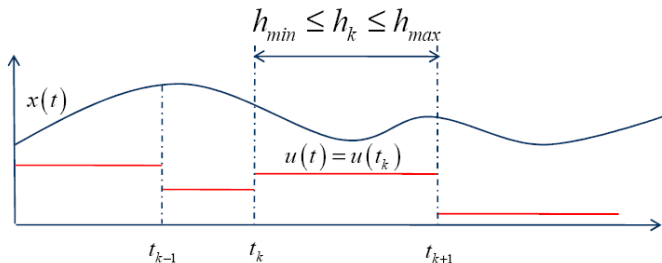


Real-time problem : the system is affected by **timing problems**

- ▶ sampling jitter (sensor, multitasking processors, packet dropouts in communication channels)
- ▶ unknown time varying delays (not adressed here)

(Wittenmark, Nilsson, Torngren, 1995)

Problem Formulation



Consider the system

$$\dot{x}(t) = Ax(t) + Bu(t), \quad \forall t \in \mathbb{R}^+$$

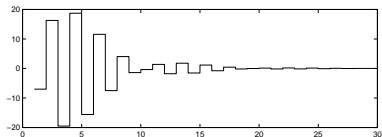
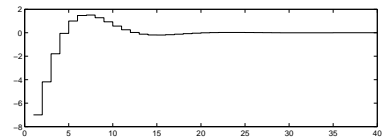
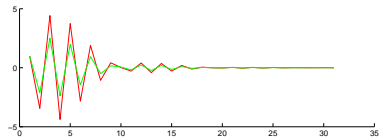
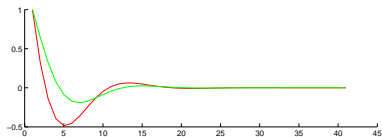
with a sampled-data control :

$$u(t) = Kx(t_k), \quad \forall t \in [t_k, t_{k+1})$$

Problem : is the system robust to jitter ?

Sampling jitter example (Zhang,2001)

$$\dot{x} = Ax + Bu_k, \quad u_k = Kx_k, \quad h_k \in \{T_1, T_2\}$$



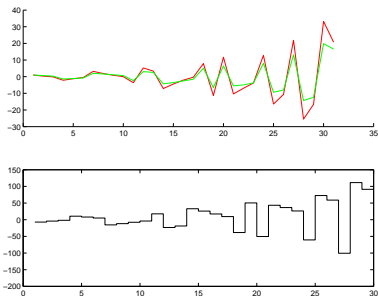
$h = 0.18s$

$h = 0.58s$

$$A = \begin{bmatrix} 1 & 3 \\ 2 & 1 \end{bmatrix} \quad B = \begin{bmatrix} 1 \\ 0.6 \end{bmatrix}; \quad K = -[1 \quad 6]$$

Sampling jitter example (Zhang,2001) \Rightarrow instability

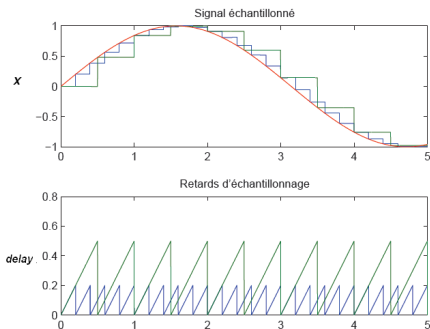
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Open problem : provide tools for robust stability and performance analysis!

Existing work - Continuous-time : time delay approach



$$u(t) = Kx(t_k) = Kx(t - \tau) \text{ with } \tau = t - t_k, 0 < \tau_k < h_{max}$$

Existing work - Continuous-time

- ▶ Fridman et al, 2004 (input delay approach)
- ▶ Mirkin, 2007 (robust control equivalent)
- ▶ Hespanha, 2008 (impulsive delay diff. eq.)

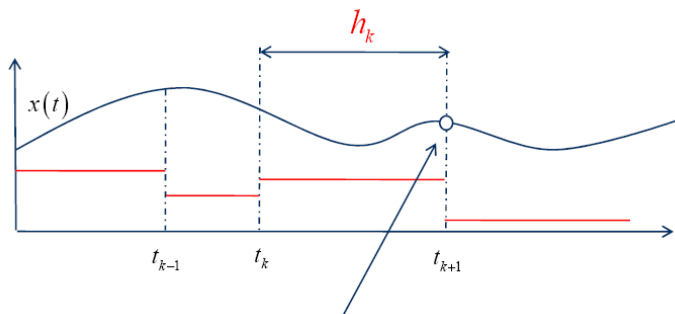
Advantage :

- ▶ Directly extend to performance study (decay rate)
- ▶ Take into account the inter-sampling behaviour

Inconvenient :

- ▶ Do not take into account the sawtooth form of the delay (conservatism)

Existing work - Discrete-time (LPV model)



$$x(t_{k+1}) = e^{(t_{k+1}-t_k)A} x(t_k) + \int_0^{(t_{k+1}-t_k)} e^{sA} ds Bu(t_{k+1})$$

$$\Rightarrow x_{k+1} = \Lambda(h_k) x(t_k)$$

Existing work - Discrete-time (LPV model)

- ▶ Sala, 2004 ; Boyd,2008 (gridding approach)
- ▶ Hetel, 2006, 2009 (convex embedding)
- ▶ Fujioka, 2007 (gridding + robust control)

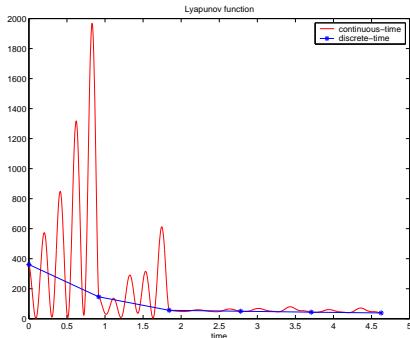
Advantage :

- ▶ Implicitly take into account the sawtooth form of the delay (less conservative)

Inconvenient :

- ▶ Not numerically efficient when the minimum sampling interval goes to zero
- ▶ Ignore the inter-sampling behaviour

Discrete-time problem (inter-sampling behaviour)



Evolution of a Lyapunov function $V(x) = x^T P x$

- ▶ Strictly decreasing at $t = t_k$
(sufficient condition for stability analysis)
- ▶ Increasing in between the sampling times
(false evaluation of control performance)

Goal

Provide a continuous-time method that takes into account the
sawtooth form of the delay
(advantage of discrete-time methods for conservatism reduction)

Exponential Stability

- ▶ how fast the norm of the state vector converges to zero. $\alpha > 0$

$$\|x(t)\| \leq e^{-\alpha t} c \|x(0)\|, \forall t > 0$$

- ▶ for a given candidate Lyapunov function $V(x)$, if

$$\frac{dV(x)}{dt} < -2\alpha V(x), \quad \forall x \neq 0$$

Case of quadratic Lyapunov functions

For $\frac{dx(t)}{dt} = Ax(t) + BKx(t - \tau(t))$, $\tau(t) := t - t_k$, $\forall t \in [t_k, t_{k+1})$

and $V(x) = x^T P x$

- ▶ Derivative of Lyapunov function

$$\frac{dV(x)}{dt} = 2x^T(t)P(Ax(t) + BKx(t - \tau)).$$

- ▶ Sawtooth evolution of delay can be introduced by using the integration operator $\Lambda(\cdot)$ used for the discrete-time model :

$$x(t) = \Lambda(t - t_k)x(t_k), \quad \forall t \in [t_k, t_{k+1})$$

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$$x(t) = \Lambda(\tau)x(t - \tau), \quad \forall t \in [t_k, t_{k+1})$$

Estimation of the decay rate

$$\alpha > 0, P \succ 0$$

$$\begin{aligned} & (K^T B^T + \Lambda^T(\tau) A^T) P \Lambda(\tau) + \Lambda^T(\tau) P (A \Lambda(\tau) + B K) \\ & \prec -2\alpha \Lambda^T(\tau) P \Lambda(\tau), \end{aligned}$$

$$\forall \tau \in [0, h_{max}).$$

Estimation of the decay rate

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- ▶ Problem : **infinite number of LMI conditions !**

Estimation of the decay rate

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$$\forall \tau \in [0, h_{max}).$$

- ▶ Problem : **infinite number of LMI conditions !**
- ▶ Solution : convex embedding with finite number of generators

Exponential uncertainty

- ▶ Integration operator

$$\begin{aligned}\Lambda(\rho) &= e^{\rho A} + \int_0^{\rho} e^{sA} ds BK \\ &= I + \int_0^{\rho} e^{sA} ds (A + BK)\end{aligned}$$

- ▶ Exponential uncertainty :

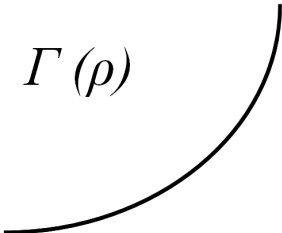
$$\Gamma(\rho) = \int_0^{\rho} e^{sA} ds, \quad \rho \in [0, h_{max}).$$

- ▶ Use a classical representation from the robust control point of view !

Exponential uncertainty - Parametric uncertainty representation

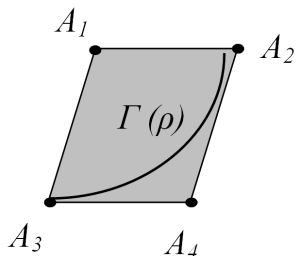
$$\Gamma(\rho) = \int_0^{\rho(k)} e^{As} ds$$
$$0 < \rho < h_{max}$$

$\Gamma(\rho)$



Treat the term like a parametric uncertainty

Exponential uncertainty - Convex Polytope



$$\Gamma(\rho) = \sum_{i=1}^N \mu_i A_i$$

$$\sum_{i=1}^N \mu_i = 1, \mu_i > 0, \forall i = 1, \dots, N$$

(Hetel, Trans. Autom. Contr. 2006)

(Cloosterman, Trans. Autom. Contr. 2009),

(Olaru, IFAC World Congress 2007),

Jordan normal form

$$\lambda_1 = -1.5, \lambda_2 = 0.2$$

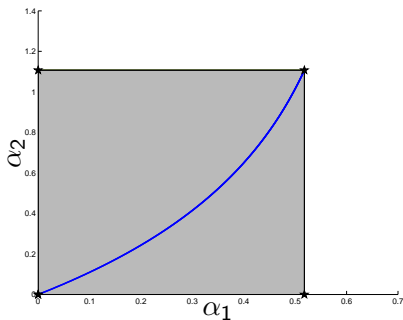
$$\text{For } A = \begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix}$$

$$\Gamma(\rho) = \begin{pmatrix} \alpha_1(\rho) & 0 \\ 0 & \alpha_2(\rho) \end{pmatrix}$$

with

$$\alpha_i(\rho) = \int_0^\rho e^{\lambda_i s} ds$$

vertex = max or min $\alpha_i(\rho)$



$$\Gamma(\rho) = \sum_{j=0}^{2^n} \mu_j A_j$$

Jordan normal form + gridding

$$\lambda_1 = -1.5, \lambda_2 = 0.2$$

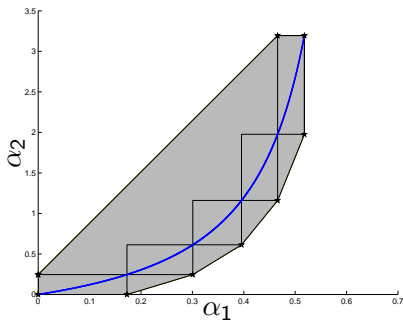
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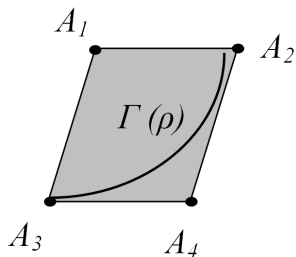
$$\alpha_i(\rho) = \int_0^\rho e^{\lambda_i s} ds$$

vertex = max or min $\alpha_i(\rho)$



$$\Gamma(\rho) = \sum_{j=0}^{5 \times 2^n} \mu_j A_j$$

Tractable LMI conditions



$$\Gamma(\rho) = \sum_{i=1}^N \mu_i A_i$$

$$\sum_{j=1}^N \mu_j = 1, \mu_j > 0, \forall j = 1, \dots, N$$

Stability conditions : $P = P^T \succ 0$

$$\begin{pmatrix} A^T P + PA + G_1 + G_1^T + \alpha P & PBK - G_1 A_j + G_2^T \\ K^T B^T P - A_j^T G_1^T + G_2 & -G_2 A_j - A_j^T G_2^T \end{pmatrix} \prec \mathbf{0},$$

$$\forall j = 1, \dots, N.$$

Numerical example

Consider a continuous-time system described by :

$$A = \begin{pmatrix} 1 & 15 \\ -15 & 1 \end{pmatrix} \text{ and } B = \begin{pmatrix} 1 \\ 1 \end{pmatrix}.$$

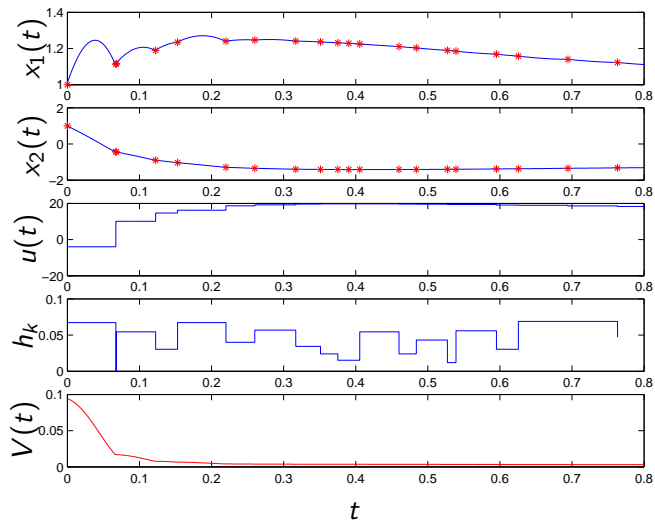
- ▶ $\lambda(A) = 1 \pm 15i$
- ▶ K - obtained by pole placement :

$$\lambda(A + BK) = -1 \pm i$$

Stability analysis comparison :

- ▶ (Mirkin, TAC 2007) : $h \in [0, 0.014]$
- ▶ (Naghshtabrizi, Hespanha, Teel, SCL 2008) : $h \in [0, 0.033]$
- ▶ (Fujioka, Automatica 2009) : $h \in [0, 0.07]$
- ▶ continuous-time convex embedding : $h \in [0, 0.09]$

Numerical example



Conclusion and Perspective

- ▶ Robustness to sampling jitter
- ▶ Provide robust methods for stability
- ▶ Show how to reduce the conservatism of stability analysis by taking into account the sawtooth form of the delay
- ▶ Perspective : control design