

Robust stability analysis of discrete-time systems  
with parametric and switching uncertainties

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To be presented at 19th IFAC World Congress / Cape Town

■ Analysis of discrete-time polytopic systems  $x_{k+1} = A(\theta_k)x_k$  where

$$A(\theta_k) = \sum_{v=1}^{\bar{v}} \theta_{k,v} A^{[v]} \quad : \quad \theta_k \in \Xi_{\bar{v}} = \left\{ \sum_{v=1}^{\bar{v}} \theta_{k,v} = 1, \theta_{k,v} \geq 0 \right\}$$

① “Quadratic stability” test [Bar85]  $\theta_{k+1} \neq \theta_k$

$$\exists P \succ 0 \quad : \quad A^{[v]T} P A^{[v]} - P \prec 0 \quad \forall v = 1 \dots \bar{v}$$

② “Slack variables” test [PABB00] assuming  $\theta_k = \theta \quad \forall k \in \mathbb{N}$

$$\exists P^{[v]} \succ 0, G \quad : \quad \begin{bmatrix} P^{[v]} & 0 \\ 0 & -P^{[v]} \end{bmatrix} \prec \left\{ G \begin{bmatrix} I & -A^{[v]} \end{bmatrix} \right\}^S \quad \forall v = 1 \dots \bar{v}$$

③ “Switching” test [DB01]  $\theta_{k+1} \neq \theta_k$

$$\exists P^{[v]} \succ 0 \quad : \quad A^{[v]T} P^{[w]} A^{[v]} - P^{[v]} \prec 0 \quad \forall v = 1 \dots \bar{v} \quad \forall w = 1 \dots \bar{v}$$

■ ②  $\Leftarrow$  ① and ③  $\Leftarrow$  ① but ②??③

- Any rationally dependent model admits an LFT representation

$$x_{k+1} = A(\theta_k)x_k \quad : \quad A(\theta) = A + B\Delta(\theta)(I - D\Delta(\theta))^{-1}C$$

where  $\Delta(\theta)$  is linear in  $\theta$ .

- Also admits a reduced size descriptor representation

$$E_x(\theta_k)x_{k+1} + E_\pi(\theta_k)\pi_k = F(\theta_k)x_k$$

where  $E_x(\theta)$ ,  $E_\pi(\theta)$ ,  $F(\theta)$  are affine in  $\theta$

- Proof: take

$$\begin{bmatrix} I \\ 0 \end{bmatrix} x_{k+1} + \begin{bmatrix} -B\Delta(\theta_k) \\ I - D\Delta(\theta_k) \end{bmatrix} \pi_k = \begin{bmatrix} A \\ C \end{bmatrix} x_k$$

- Remark:  $\begin{bmatrix} E_x(\theta) & E_\pi(\theta) \end{bmatrix}$  is full column rank if LFT is well-posed.

■ Example:  $a_k y_{k+2} + b_k^2 y_{k+1} + a_k b_k y_k = 0$

● LFT model:  $\Delta(\theta)$  is at least  $4 \times 4$  (i.e. model with 8 exogenous signals), e.g.

$$A(a, b) = \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix} + \begin{bmatrix} -1 & 0 & -1 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \Delta \left( I - \begin{bmatrix} 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 0 \\ -1 & 0 & -1 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \Delta \right)^{-1} \begin{bmatrix} 0 & 1 \\ 1 & 0 \\ 0 & 0 \\ 0 & 1 \end{bmatrix}$$

with  $\Delta = \begin{bmatrix} bI_2 & 0 \\ 0 & (a-1)I_2 \end{bmatrix}$

● Descriptor model with 1 exogenous signal

$$\begin{bmatrix} a_k & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix} x_{k+1} + \begin{bmatrix} b_k \\ 0 \\ 1 \end{bmatrix} \pi_k = \begin{bmatrix} 0 & 0 \\ 1 & 0 \\ b_k & a_k \end{bmatrix} x_k$$

- Polytopic descriptor system with uncertain and witching parameters:

$$E_x(\phi, \theta_k)x_{k+1} + E_\pi(\phi, \theta_k)\pi_k - F(\phi, \theta_k)x_k = M(\phi, \theta_k) \begin{pmatrix} x_{k+1} \\ \pi_k \\ x_k \end{pmatrix} = 0$$

$$M(\phi, \theta_k) = \sum_{\mu=1}^{\bar{\mu}} \sum_{v=1}^{\bar{v}} \phi_\mu \theta_{k,v} M^{[\mu,v]} \quad : \quad \phi \in \Xi_{\bar{\mu}} \quad , \quad \theta_k \in \Xi_{\bar{v}}$$

- ▲ System assumed well-posed:  $\begin{bmatrix} E_x(\phi, \theta) & E_\pi(\phi, \theta) \end{bmatrix}$  is strict full column rank.

- Stability is assed if  $\exists P^{[\mu,v]} \succ 0, G^{[v]}$ :

$$\begin{bmatrix} P^{[\mu,w]} & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & -P^{[\mu,v]} \end{bmatrix} \prec \left\{ G^{[w]} M^{[\mu,v]} \right\}^S \quad \begin{array}{l} \forall \mu = 1 \dots \bar{\mu} \\ \forall v = 1 \dots \bar{v} \\ \forall w = 1 \dots \bar{v} \end{array}$$

- Contains all previous results as special cases. High numerical complexity.

$$\begin{bmatrix} P^{[\mu,w]} & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & -P^{[\mu,v]} \end{bmatrix} \prec \left\{ G^{[w]} M^{[\mu,v]} \right\}^S \quad \begin{array}{l} \forall \mu = 1 \dots \bar{\mu} \\ \forall v = 1 \dots \bar{v} \\ \forall w = 1 \dots \bar{w} \end{array}$$

■ Numerical complexity can be reduced if

- ▲ some of the columns of  $\begin{bmatrix} E_x(\phi, \theta) & E_\pi(\phi, \theta) \end{bmatrix}$  are independent of  $\phi, \theta$
- ▲ or some of the rows of  $M(\phi, \theta)$  are independent of  $\phi, \theta$ .

● Example of the switching ordinary system  $M(\theta) = \begin{bmatrix} I & -A(\theta) \end{bmatrix}$

▲ lossless reduced size LMIs (no need for slack variables)

$$A^{[v]T} P^{[w]} A^{[v]} - P^{[v]} \prec 0 \quad \forall v = 1 \dots \bar{v} \quad \forall w = 1 \dots \bar{w}$$

- Large size LMI for the numerical example

$$\mathcal{P}(P^{[\mu,w]}, P^{[\mu,v]}) \prec \left\{ G^{[w]} \left[ \begin{array}{cc|c|cc} a^{[\bullet]} & 0 & b^{[\star]} & 0 & 0 \\ 0 & 1 & 0 & -1 & 0 \\ 0 & 0 & 1 & -b^{[\star]} & -a^{[\bullet]} \end{array} \right] \right\}^S$$

- lossless reduced size LMIs (both in nb of rows and nb of vars)

$$\hat{\mathcal{P}}(P^{[\mu,w]}, P^{[\mu,v]}) \prec \left\{ \hat{G}^{[w]} \left[ \begin{array}{cccc} a^{[\bullet]} & 0 & b^{[\star]} & 0 \\ 0 & -b^{[\star]} & 1 & -a^{[\bullet]} \end{array} \right] \right\}^S$$

▲ ● =  $\mu$  if  $a$  is parametric (constant) and ● =  $v$  if  $a_k$  is switching

▲ ★ =  $\mu$  if  $b$  is parametric (constant) and ★ =  $v$  if  $b_k$  is switching

- Numerical results for  $a \in [1, 2]$  and  $b \in [-0.5, \beta]$ .
- Goal: find maximal  $\beta$  such that LMIs are feasible

	$\beta$ (nb vars/nb rows)
$a_k, b_k$	0.81094 (44/64)
$a, b_k$	0.89027 (28/32)
$a_k, b$	0.82658 (28/32)
$a, b$	0.98059 (20/16)



- Considered example, model before augmentation:

$$a_k y_{k+2} + b_k^2 y_{k+1} + a_k b_k y_k = 0$$

- After one step ahead augmentation:

$$\begin{cases} a_k y_{k+2} + b_k^2 y_{k+1} + a_k b_k y_k = 0 \\ a_{k+1} y_{k+3} + b_{k+1}^2 y_{k+2} + a_{k+1} b_{k+1} y_{k+1} = 0 \end{cases}$$

- Corresponding descriptor representation

$$\begin{bmatrix} a_{k+1} & 0 & 0 \\ 0 & a_k & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \tilde{x}_{k+1} + \begin{bmatrix} b_{k+1} & 0 \\ 0 & b_k \\ 0 & 0 \\ 0 & 0 \\ 1 & 0 \\ 0 & 1 \end{bmatrix} \tilde{\pi}_k = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \\ b_{k+1} & a_{k+1} & 0 \\ 0 & b_k & a_k \end{bmatrix} \tilde{x}_k$$

- Results applied to augmented model are guaranteed to be less conservative

■ Numerical results for  $a \in [1, 2]$  and  $b \in [-0.5, \beta]$ .

● Goal: find maximal  $\beta$  such that LMIs are feasible

$\beta$ (nb vars/nb rows)	original model	augmented	upper limit
$a_k, b_k$	0.81094 (44/64)	0.84677 (480/1536)	?
$a, b_k$	0.89027 (28/32)	0.90293 (144/192)	?
$a_k, b$	0.82658 (28/32)	0.85375 (144/192)	?
$a, b$	0.98059 (20/16)	0.99519 (48/24)	1

## ■ New LMI test

- Combines existing techniques w.r.t. uncertain & switching parameters
- Extends results to descriptor-like models (represent any rationally-dependent system)
- Contributes to conservatism reduction thanks to the system augmentation technique
- ▲ Numerical complexity is increased ... but can be controlled

## ■ Prospective work

- Continuous time case
- Time-varying parameters with bounded rate
- Extension to design problems (state & output feedback, robust observers...)

## ■ Springer monograph by Y. Ebihara & D. Peaucelle to be published in 2014-2015

## References

- [Bar85] B.R. Barmish, *Necessary and sufficient condition for quadratic stabilizability of an uncertain system*, J. Optimization Theory and Applications **46** (1985), no. 4.
- [DB01] J. Daafouz and J. Bernussou, *Parameter dependent lyapunov functions for discrete time systems with time varying parametric uncertainties*, Systems & Control Letters **43** (2001), 355–359.
- [PABB00] D. Peaucelle, D. Arzelier, O. Bachelier, and J. Bernussou, *A new robust D-stability condition for real convex polytopic uncertainty*, Systems & Control Letters **40** (2000), no. 1, 21–30.