

# Analysis of control of LPV systems with piecewise constant parameters

Corentin Briat  
ETH-Zürich - D-BSSE

MOSAR Workshop  
Onera, Toulouse, France - 28/11/2014

**ETH**

Eidgenössische Technische Hochschule Zürich  
Swiss Federal Institute of Technology Zurich



**D-BSSE**

Department of Biosystems  
Science and Engineering

**C·T·S·B**  
CONTROL THEORY & SYSTEMS BIOLOGY



# Introduction



## LPV systems

$$\begin{aligned}\dot{x}(t) &= A(\rho(t))x(t) + B(\rho(t))u(t) \\ y(t) &= C(\rho(t))x(t) + D(\rho(t))u(t) \\ x(0) &= x_0\end{aligned}\tag{1}$$

where

- $x$ ,  $u$  and  $y$  are the state of the system, the (control) input and the output
- $\rho$  is the parameter vector
- Matrix-valued functions  $A(\cdot)$ ,  $B(\cdot)$ ,  $C(\cdot)$  and  $D(\cdot)$  “nice enough”



## LPV systems

$$\begin{aligned}
 \dot{x}(t) &= A(\rho(t))x(t) + B(\rho(t))u(t) \\
 y(t) &= C(\rho(t))x(t) + D(\rho(t))u(t) \\
 x(0) &= x_0
 \end{aligned} \tag{1}$$

where

- $x$ ,  $u$  and  $y$  are the state of the system, the (control) input and the output
- $\rho$  is the parameter vector
- Matrix-valued functions  $A(\cdot)$ ,  $B(\cdot)$ ,  $C(\cdot)$  and  $D(\cdot)$  "nice enough"

## Commonly considered parameters

- Bounded differentiable trajectories
- Discontinuous bounded trajectories
- Periodic, switched and Markov jump systems can also be seen as LPV systems



# What is the rationale for LPV systems?

## Stability analysis

- Identical to uncertain time-varying parametric systems
- LPV analysis = robust analysis



# What is the rationale for LPV systems?

## Stability analysis

- Identical to uncertain time-varying parametric systems
- LPV analysis = robust analysis

## LPV design

- We assume in the LPV framework that the parameters are measured/known
- **So, we can use them in controllers, observers, etc.**
- For instance, a gain-scheduled state-feedback controller would take the form

$$u(t) = K(\rho(t))x(t) \quad (2)$$

- LPV control > robust control
- **But LPV controllers are more difficult to design!**



D-BSSE  
Department of Biosystems  
Science and Engineering

# Stability analysis of LPV systems



# Quadratic stability

## Theorem

*The LPV system*

$$\begin{aligned}
 \dot{x}(t) &= A(\rho(t))x(t) \\
 x(0) &= x_0
 \end{aligned}
 \tag{3}$$

*is quadratically stable if and only if there exists a matrix  $P \in \mathbb{S}_{>0}^n$  such that the LMI*

$$A(\theta)^T P + P A(\theta) \prec 0
 \tag{4}$$

*holds for all  $\theta \in \mathcal{P}$ .*







## Theorem

The LPV system

$$\begin{aligned}\dot{x}(t) &= A(\rho(t))x(t) \\ x(0) &= x_0\end{aligned}\tag{5}$$

with  $\rho \in \{f : \mathbb{R}_{\geq 0} \rightarrow \mathcal{P} \subset \mathbb{R}^N, f'(t) \in \mathcal{D}, t \geq 0\}$  is robustly stable if and only if there exists a differentiable matrix-valued function  $P : \mathcal{P} \rightarrow \mathbb{S}_{>0}^n$  such that the LMI

$$\sum_{i=1}^N \theta'_i \frac{\partial}{\partial \theta_i} P(\theta) + A(\theta)^T P(\theta) + P(\theta) A(\theta) \prec 0\tag{6}$$

holds for all  $\theta \in \mathcal{P}$  and all  $\theta' \in \mathcal{D}$ .



## Theorem

The LPV system

$$\begin{aligned}\dot{x}(t) &= A(\rho(t))x(t) \\ x(0) &= x_0\end{aligned}\tag{5}$$

with  $\rho \in \{f : \mathbb{R}_{\geq 0} \rightarrow \mathcal{P} \subset \mathbb{R}^N, f'(t) \in \mathcal{D}, t \geq 0\}$  is robustly stable if and only if there exists a differentiable matrix-valued function  $P : \mathcal{P} \rightarrow \mathbb{S}_{>0}^n$  such that the LMI

$$\sum_{i=1}^N \theta'_i \frac{\partial}{\partial \theta_i} P(\theta) + A(\theta)^T P(\theta) + P(\theta) A(\theta) \prec 0\tag{6}$$

holds for all  $\theta \in \mathcal{P}$  and all  $\theta' \in \mathcal{D}$ .

## Remarks

- Parameter-dependent Lyapunov function  $V(x) = x^T P(\rho)x$
- Trajectories of the parameters are continuously differentiable
- $A(\rho)$  Hurwitz stable for all  $\rho \in \mathcal{P}$  is necessary and sufficient for robust stability provided that the parameters vary sufficiently slowly
- Infinite-dimensional LMI problem



## Remarks on LPV systems

- Two main classes of parameter trajectories associated with two main stability concepts
- Quadratic stability may be conservative while robust stability too demanding
- Part of the success of periodic, switched and jump systems lies in the “tailoredness” of the tools
- The definition of the parameter trajectories is way too loose to lead to accurate results (e.g. asymptotic stability does not imply quadratic stability)



## Remarks on LPV systems

- Two main classes of parameter trajectories associated with two main stability concepts
- Quadratic stability may be conservative while robust stability too demanding
- Part of the success of periodic, switched and jump systems lies in the “tailoredness” of the tools
- The definition of the parameter trajectories is way too loose to lead to accurate results (e.g. asymptotic stability does not imply quadratic stability)

## Proposal

- What about something in between the set of all possible trajectories and those that are continuously differentiable?
- For instance, we can consider piecewise continuous/constant parameter trajectories
- Quadratic and robust stability not adapted
- Need something new!



# Stability analysis of LPV systems with piecewise constant parameters





## Two main class of parameters

- Periodic changes  $\rightarrow$  constant dwell-time
- Aperiodic changes  $\rightarrow$  minimum dwell-time

## Stability results

- Discrete-time-like stability conditions
- Lifted conditions











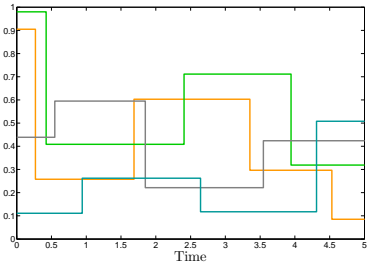
# Stability under minimum dwell-time

Let us consider the LPV system

$$\dot{x} = A(\rho)x, \quad x(0) = x_0 \tag{10}$$

with piecewise constant parameter  $\rho \in \mathcal{P}_{\geq \bar{T}}$  where

$$\mathcal{P}_{\geq \bar{T}} = \left\{ \rho : \mathbb{R}_{\geq 0} \rightarrow \mathcal{P} : \rho(t) = \rho(t_k), t \in [t_k, t_{k+1}) \right. \\ \left. t_{k+1} - t_k \geq \bar{T}, k \in \mathbb{N}_0 \right\}. \tag{11}$$



**ETH** Zürich

















# Lifted conditions - Constant DT

## Theorem

The following statements are equivalent:

- (a) There exists a matrix-valued function  $P : \mathcal{P} \rightarrow \mathbb{S}_{>0}^n$  such that the condition

$$e^{A(\theta)^T \bar{T}} P(\theta) e^{A(\theta) \bar{T}} - P(\eta) \prec 0 \tag{14}$$

holds for all  $\theta, \eta \in \mathcal{P}$ .

- (b) There exists a matrix-valued function  $S : [0, \bar{T}] \times \mathcal{P} \rightarrow \mathbb{S}^n$ ,  $S(\bar{T}, \theta) \succ 0$ , such that the conditions

$$\partial_\tau S(\tau, \theta) + \text{Sym}[S(\tau, \theta)A(\theta)] \preceq 0 \tag{15}$$

and

$$S(0, \theta) - S(\bar{T}, \eta) \prec 0 \tag{16}$$

hold for all  $\theta, \eta \in \mathcal{P}$  and all  $\tau \in \mathcal{T} := [0, \bar{T}]$ .

Moreover, when one of the above statements holds, the LPV system with piecewise constant parameters and constant dwell-time  $\bar{T}$  is asymptotically stable.









# Lifted conditions - Minimum DT

## Theorem

The following statements are equivalent:

- (a) There exists a matrix-valued function  $P : \mathcal{P} \rightarrow \mathbb{S}_{>0}^n$  such that the conditions

$$A(\theta)^T P(\theta) + P(\theta) A(\theta) \prec 0 \tag{18}$$

$$e^{A(\theta)^T \bar{T}} P(\theta) e^{A(\theta) \bar{T}} - P(\eta) \prec 0 \tag{19}$$

hold for all  $\theta, \eta \in \mathcal{P}$ .



# Lifted conditions - Minimum DT

## Theorem

The following statements are equivalent:

(a) There exists a matrix-valued function  $P : \mathcal{P} \rightarrow \mathbb{S}_{>0}^n$  such that the conditions

$$A(\theta)^T P(\theta) + P(\theta) A(\theta) \prec 0 \tag{18}$$

$$e^{A(\theta)^T \bar{T}} P(\theta) e^{A(\theta) \bar{T}} - P(\eta) \prec 0 \tag{19}$$

hold for all  $\theta, \eta \in \mathcal{P}$ .

(b) There exists a matrix-valued function  $S : [0, \bar{T}] \times \mathcal{P} \rightarrow \mathbb{S}^n$ ,  $S(\bar{T}, \theta) \succ 0$ , such that the conditions

$$A(\theta)^T S(\bar{T}, \theta) + S(\bar{T}, \theta) A(\theta) \prec 0 \tag{20}$$

$$\partial_\tau S(\tau, \theta) + \text{Sym}[S(\tau, \theta) A(\theta)] \preceq 0 \tag{21}$$

$$S(0, \theta) - S(\bar{T}, \eta) \prec 0 \tag{22}$$

hold for all  $\theta, \eta \in \mathcal{P}$  and all  $\tau \in \mathcal{T} := [0, \bar{T}]$ .







## Connection with quadratic and robust stability

### Theorem (Quadratic stability)

When  $\bar{T} \rightarrow 0$  in the minimum dwell-time theorem, then the quadratic stability condition

$$A(\theta)^T P + P A(\theta) \prec 0 \quad (23)$$

is recovered.





## Connection with quadratic and robust stability

### Theorem (Quadratic stability)

When  $\bar{T} \rightarrow 0$  in the minimum dwell-time theorem, then the quadratic stability condition

$$A(\theta)^T P + P A(\theta) \prec 0 \quad (23)$$

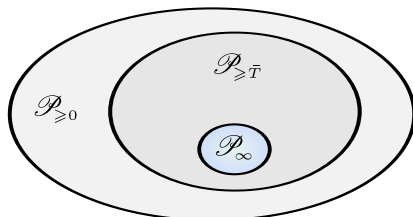
is recovered.

### Theorem (Robust stability)

When  $\bar{T} \rightarrow \infty$  in the minimum dwell-time theorem, then the robust stability condition

$$A(\theta)^T P(\theta) + P(\theta) A(\theta) \prec 0 \quad (24)$$

for constant parametric uncertainties is recovered.





# Connection with switched systems

## Switched systems

Let  $\mathcal{P} = \{1, \dots, M\}$ , for some finite  $M \in \mathbb{N}$ , and define

$$A(\rho) = \sum_{i=1}^M \delta_{i,\rho} A_i \tag{25}$$

where  $\delta_{i,j}$  is the Kronecker delta; i.e.  $\delta_{i,j} = 1$  if  $i = j$ , and 0 otherwise.

**ETH** Zürich







# Computational aspects - One parameter case

## Preliminaries

- The sets  $\mathcal{P}$  and  $\mathcal{T}$  are defined as

$$\begin{aligned}\mathcal{P} &:= \{\theta \in \mathbb{R} : g(\theta) := (\theta_{max} - \theta)(\theta - \theta_{min}) \geq 0\} \\ \mathcal{T} &:= \{\tau \in \mathbb{R} : h(\tau) := \tau(\bar{T} - \tau) \geq 0\}\end{aligned}$$

- We say that that a symmetric matrix-valued function  $M(\cdot)$  is a matrix sum of squares if there exists a matrix-valued function  $N(\cdot)$  such that  $M(\cdot) = N(\cdot)^T N(\cdot)$ .



# Computational aspects - One parameter case

## Preliminaries

- The sets  $\mathcal{P}$  and  $\mathcal{T}$  are defined as

$$\begin{aligned}
 \mathcal{P} &:= \{ \theta \in \mathbb{R} : g(\theta) := (\theta_{max} - \theta)(\theta - \theta_{min}) \geq 0 \} \\
 \mathcal{T} &:= \{ \tau \in \mathbb{R} : h(\tau) := \tau(\bar{T} - \tau) \geq 0 \}
 \end{aligned}$$

- We say that that a symmetric matrix-valued function  $M(\cdot)$  is a matrix sum of squares if there exists a matrix-valued function  $N(\cdot)$  such that  $M(\cdot) = N(\cdot)^T N(\cdot)$ .

## Proposition

Let  $\varepsilon_1, \varepsilon_2, \bar{T} > 0$  be given and assume that there exist polynomial matrix-valued functions  $S, \Gamma_j : \mathbb{R}^2 \rightarrow \mathbb{S}^n, j = 1, \dots, 4$  and  $\Gamma : \mathbb{R} \rightarrow \mathbb{S}^n$  such that

- $\Gamma, \Gamma_j, j = 1, \dots, 4$ , are SOS matrix polynomials





# Computational aspects - One parameter case

## Preliminaries

- The sets  $\mathcal{P}$  and  $\mathcal{T}$  are defined as

$$\begin{aligned}
 \mathcal{P} &:= \{ \theta \in \mathbb{R} : g(\theta) := (\theta_{max} - \theta)(\theta - \theta_{min}) \geq 0 \} \\
 \mathcal{T} &:= \{ \tau \in \mathbb{R} : h(\tau) := \tau(\bar{T} - \tau) \geq 0 \}
 \end{aligned}$$

- We say that that a symmetric matrix-valued function  $M(\cdot)$  is a matrix sum of squares if there exists a matrix-valued function  $N(\cdot)$  such that  $M(\cdot) = N(\cdot)^T N(\cdot)$ .

## Proposition

Let  $\varepsilon_1, \varepsilon_2, \bar{T} > 0$  be given and assume that there exist polynomial matrix-valued functions  $S, \Gamma_j : \mathbb{R}^2 \rightarrow \mathbb{S}^n, j = 1, \dots, 4$  and  $\Gamma : \mathbb{R} \rightarrow \mathbb{S}^n$  such that

- $\Gamma, \Gamma_j, j = 1, \dots, 4$ , are SOS matrix polynomials
- $S(\bar{T}, \theta) - \Gamma(\theta)g(\theta) - \varepsilon_1 I_n$  is SOS

ETH zürich





# Computational aspects - One parameter case

## Preliminaries

- The sets  $\mathcal{P}$  and  $\mathcal{T}$  are defined as

$$\begin{aligned}\mathcal{P} &:= \{\theta \in \mathbb{R} : g(\theta) := (\theta_{max} - \theta)(\theta - \theta_{min}) \geq 0\} \\ \mathcal{T} &:= \{\tau \in \mathbb{R} : h(\tau) := \tau(\bar{T} - \tau) \geq 0\}\end{aligned}$$

- We say that a symmetric matrix-valued function  $M(\cdot)$  is a matrix sum of squares if there exists a matrix-valued function  $N(\cdot)$  such that  $M(\cdot) = N(\cdot)^T N(\cdot)$ .

## Proposition

Let  $\varepsilon_1, \varepsilon_2, \bar{T} > 0$  be given and assume that there exist polynomial matrix-valued functions  $S, \Gamma_j : \mathbb{R}^2 \rightarrow \mathbb{S}^n, j = 1, \dots, 4$  and  $\Gamma : \mathbb{R} \rightarrow \mathbb{S}^n$  such that

- $\Gamma, \Gamma_j, j = 1, \dots, 4$ , are SOS matrix polynomials
- $S(\bar{T}, \theta) - \Gamma(\theta)g(\theta) - \varepsilon_1 I_n$  is SOS
- $-\partial_\tau S(\tau, \theta) - \text{Sym}[S(\tau, \theta)A(\theta)] - \Gamma_1(\tau, \theta)h(\tau) - \Gamma_2(\tau, \theta)g(\theta)$  is SOS



# Computational aspects - One parameter case

## Preliminaries

- The sets  $\mathcal{P}$  and  $\mathcal{T}$  are defined as

$$\begin{aligned}\mathcal{P} &:= \{ \theta \in \mathbb{R} : g(\theta) := (\theta_{max} - \theta)(\theta - \theta_{min}) \geq 0 \} \\ \mathcal{T} &:= \{ \tau \in \mathbb{R} : h(\tau) := \tau(\bar{T} - \tau) \geq 0 \}\end{aligned}$$

- We say that a symmetric matrix-valued function  $M(\cdot)$  is a matrix sum of squares if there exists a matrix-valued function  $N(\cdot)$  such that  $M(\cdot) = N(\cdot)^T N(\cdot)$ .

## Proposition

Let  $\varepsilon_1, \varepsilon_2, \bar{T} > 0$  be given and assume that there exist polynomial matrix-valued functions  $S, \Gamma_j : \mathbb{R}^2 \rightarrow \mathbb{S}^n, j = 1, \dots, 4$  and  $\Gamma : \mathbb{R} \rightarrow \mathbb{S}^n$  such that

- $\Gamma, \Gamma_j, j = 1, \dots, 4$ , are SOS matrix polynomials
- $S(\bar{T}, \theta) - \Gamma(\theta)g(\theta) - \varepsilon_1 I_n$  is SOS
- $-\partial_\tau S(\tau, \theta) - \text{Sym}[S(\tau, \theta)A(\theta)] - \Gamma_1(\tau, \theta)h(\tau) - \Gamma_2(\tau, \theta)g(\theta)$  is SOS
- $S(\bar{T}, \eta) - S(0, \theta) - \varepsilon_2 I - \Gamma_3(\theta, \eta)g(\theta) - \Gamma_4(\theta, \eta)g(\eta)$  is SOS



## Computational aspects - One parameter case

### Preliminaries

- The sets  $\mathcal{P}$  and  $\mathcal{T}$  are defined as

$$\begin{aligned}\mathcal{P} &:= \{\theta \in \mathbb{R} : g(\theta) := (\theta_{max} - \theta)(\theta - \theta_{min}) \geq 0\} \\ \mathcal{T} &:= \{\tau \in \mathbb{R} : h(\tau) := \tau(\bar{T} - \tau) \geq 0\}\end{aligned}$$

- We say that a symmetric matrix-valued function  $M(\cdot)$  is a matrix sum of squares if there exists a matrix-valued function  $N(\cdot)$  such that  $M(\cdot) = N(\cdot)^T N(\cdot)$ .

### Proposition

Let  $\varepsilon_1, \varepsilon_2, \bar{T} > 0$  be given and assume that there exist polynomial matrix-valued functions  $S, \Gamma_j : \mathbb{R}^2 \rightarrow \mathbb{S}^n, j = 1, \dots, 4$  and  $\Gamma : \mathbb{R} \rightarrow \mathbb{S}^n$  such that

- $\Gamma, \Gamma_j, j = 1, \dots, 4$ , are SOS matrix polynomials
- $S(\bar{T}, \theta) - \Gamma(\theta)g(\theta) - \varepsilon_1 I_n$  is SOS
- $-\partial_\tau S(\tau, \theta) - \text{Sym}[S(\tau, \theta)A(\theta)] - \Gamma_1(\tau, \theta)h(\tau) - \Gamma_2(\tau, \theta)g(\theta)$  is SOS
- $S(\bar{T}, \eta) - S(0, \theta) - \varepsilon_2 I - \Gamma_3(\theta, \eta)g(\theta) - \Gamma_4(\theta, \eta)g(\eta)$  is SOS

Then the LPV system with piecewise constant parameters and constant dwell-time  $\bar{T}$  is asymptotically stable.



## Example

- Let us consider here an LPV system with the matrix

$$A(\theta) = \begin{bmatrix} 0 & 1 \\ -2 - \theta & -1 \end{bmatrix} \quad (28)$$

where  $\theta \in [0, \bar{\theta}]$ ,  $\bar{\theta} > 0$ .

- It is known that this system is quadratically stable if and only if  $\bar{\theta} \leq 3.828$ .





# Control of LPV systems with piecewise constant parameters



## System

- Let us consider the LPV system

$$\begin{aligned} \dot{x}(t) &= A(\rho(t))x(t) + B(\rho(t))u(t) \\ x(0) &= x_0 \end{aligned}$$

where  $\{t_k\}_{k \in \mathbb{N}_0}$  is the sequence of time instants at which the parameter vector changes value.











## State-feedback control - Constant DT

### Theorem

The following statements are equivalent:

- (a) There exists a matrix-valued function  $P : \mathcal{P} \rightarrow \mathbb{S}_{>0}^n$  such that the condition

$$\Phi_{\theta}(\bar{T})^T P(\theta) \Phi_{\theta}(\bar{T}) - P(\eta) \prec 0 \quad (31)$$

holds for all  $\theta, \eta \in \mathcal{P}$  where

$$\Phi'_{\theta}(s) = (A(\theta) + B(\theta)K(s, \theta))\Phi_{\theta}(s), \quad \Phi_{\theta}(0) = I, \quad s \in [0, \bar{T}]. \quad (32)$$



# State-feedback control - Constant DT

## Theorem

The following statements are equivalent:

- (a) *There exists a matrix-valued function  $P : \mathcal{P} \rightarrow \mathbb{S}_{>0}^n$  such that the condition*

$$\Phi_\theta(\bar{T})^T P(\theta) \Phi_\theta(\bar{T}) - P(\eta) \prec 0 \tag{31}$$

*holds for all  $\theta, \eta \in \mathcal{P}$  where*

$$\Phi'_\theta(s) = (A(\theta) + B(\theta)K(s, \theta))\Phi_\theta(s), \quad \Phi_\theta(0) = I, \quad s \in [0, \bar{T}]. \tag{32}$$

- (b) *There exists a matrix-valued function  $\tilde{S} : [0, \bar{T}] \times \mathcal{P} \rightarrow \mathbb{S}^n$ ,  $\tilde{S}(\bar{T}, \theta) \succ 0$ , such that the conditions*

$$-\partial_\tau \tilde{S}(\tau, \theta) + \text{Sym}[A(\theta)\tilde{S}(\tau, \theta) + B(\theta)U(\tau, \theta)] \preceq 0 \tag{33}$$

*and*

$$\tilde{S}(\bar{T}, \eta) - \tilde{S}(0, \theta) \prec 0 \tag{34}$$

*hold for all  $\theta, \eta \in \mathcal{P}$  and all  $\tau \in [0, \bar{T}]$ .*



# State-feedback control - Constant DT

## Theorem

The following statements are equivalent:

- (a) There exists a matrix-valued function  $P : \mathcal{P} \rightarrow \mathbb{S}_{>0}^n$  such that the condition

$$\Phi_\theta(\bar{T})^T P(\theta) \Phi_\theta(\bar{T}) - P(\eta) \prec 0 \tag{31}$$

holds for all  $\theta, \eta \in \mathcal{P}$  where

$$\Phi'_\theta(s) = (A(\theta) + B(\theta)K(s, \theta))\Phi_\theta(s), \quad \Phi_\theta(0) = I, \quad s \in [0, \bar{T}]. \tag{32}$$

- (b) There exists a matrix-valued function  $\tilde{S} : [0, \bar{T}] \times \mathcal{P} \rightarrow \mathbb{S}^n$ ,  $\tilde{S}(\bar{T}, \theta) \succ 0$ , such that the conditions

$$-\partial_\tau \tilde{S}(\tau, \theta) + \text{Sym}[A(\theta)\tilde{S}(\tau, \theta) + B(\theta)U(\tau, \theta)] \preceq 0 \tag{33}$$

and

$$\tilde{S}(\bar{T}, \eta) - \tilde{S}(0, \theta) \prec 0 \tag{34}$$

hold for all  $\theta, \eta \in \mathcal{P}$  and all  $\tau \in [0, \bar{T}]$ .

Moreover, when one of the above statements holds, then the closed-loop LPV system is asymptotically stable with constant dwell-time  $\bar{T}$  and a suitable controller gain can be computed using  $K(\tau, \theta) = U(\tau, \theta)\tilde{S}(\tau, \theta)^{-1}$ .



# State-feedback control - Minimum DT

## Theorem

Assume that there exists a matrix-valued function  $\tilde{S} : [0, \bar{T}] \times \mathcal{P} \rightarrow \mathbb{S}^n$ ,  $\tilde{S}(\bar{T}, \theta) \succ 0$ , such that the conditions

$$\text{Sym}[A(\theta)\tilde{S}(\bar{T}, \theta) + B(\theta)U(\bar{T}, \theta)] \prec 0, \tag{35}$$

$$-\partial_\tau \tilde{S}(\tau, \theta) + \text{Sym}[A(\theta)\tilde{S}(\tau, \theta) + B(\theta)U(\tau, \theta)] \preceq 0 \tag{36}$$

and

$$\tilde{S}(\bar{T}, \eta) - \tilde{S}(0, \theta) \prec 0 \tag{37}$$

hold for all  $\theta, \eta \in \mathcal{P}$  and all  $\tau \in [0, \bar{T}]$ .





## System

$$\dot{x} = \begin{bmatrix} 3 - \theta & 1 \\ 1 - \theta & 2 + \theta \end{bmatrix} x + \begin{bmatrix} 1 \\ 1 + \theta \end{bmatrix} u, \theta \in [0, 1]. \tag{39}$$





## System

$$\dot{x} = \begin{bmatrix} 3 - \theta & 1 \\ 1 - \theta & 2 + \theta \end{bmatrix} x + \begin{bmatrix} 1 \\ 1 + \theta \end{bmatrix} u, \theta \in [0, 1]. \tag{39}$$

## Proposition

No control law of the form  $u = K(\theta)x$  can quadratically stabilize the system (39).





## System

$$\dot{x} = \begin{bmatrix} 3 - \theta & 1 \\ 1 - \theta & 2 + \theta \end{bmatrix} x + \begin{bmatrix} 1 \\ 1 + \theta \end{bmatrix} u, \theta \in [0, 1]. \quad (39)$$

## Proposition

No control law of the form  $u = K(\theta)x$  can quadratically stabilize the system (39).

## Proof

- Quadratically stabilizable if and only if the LMI

$$L(\theta) := B_{\perp}(\theta)[A(\theta)P + PA(\theta)^T]B_{\perp}(\theta)^T \prec 0$$

is feasible for all  $\theta \in [0, 1]$  where  $B_{\perp}(\theta) = \begin{bmatrix} 1 + \theta & -1 \end{bmatrix}$ .





## System

$$\dot{x} = \begin{bmatrix} 3 - \theta & 1 \\ 1 - \theta & 2 + \theta \end{bmatrix} x + \begin{bmatrix} 1 \\ 1 + \theta \end{bmatrix} u, \theta \in [0, 1]. \quad (39)$$

## Proposition

No control law of the form  $u = K(\theta)x$  can quadratically stabilize the system (39).

## Proof

- Quadratically stabilizable if and only if the LMI

$$L(\theta) := B_{\perp}(\theta)[A(\theta)P + PA(\theta)^T]B_{\perp}(\theta)^T \prec 0$$

is feasible for all  $\theta \in [0, 1]$  where  $B_{\perp}(\theta) = \begin{bmatrix} 1 + \theta & -1 \end{bmatrix}$ .

- Assume it is stabilizable, then  $L(0) \prec 0$  and  $L(1) \prec 0$ .
- This implies that there exists a  $p \in \mathbb{R}$  such that

$$f_1(p) = p^2 - 3p + 2 < 0 \quad \text{and} \quad f_2(p) = p^2 - 6p + 8 < 0. \quad (40)$$

- But  $f_1(p) < 0 \Leftrightarrow p \in (1, 2)$  and  $f_2(p) < 0 \Leftrightarrow p \in (2, 4)$ .





# Example

- We pick  $\bar{T} = 0.05$ , polynomials  $S, U$  of order 1, and polynomials  $\Gamma$ 's of order 2.





# Example

- We pick  $\bar{T} = 0.05$ , polynomials  $S, U$  of order 1, and polynomials  $\Gamma$ 's of order 2.
- Primal/dual variables: 551/120; computation time: less than 2sec



## Example

- We pick  $\bar{T} = 0.05$ , polynomials  $S, U$  of order 1, and polynomials  $\Gamma$ 's of order 2.
- Primal/dual variables: 551/120; computation time: less than 2sec
- We find

$$K(\tau, \theta) = \frac{1}{\text{den}(\tau, \theta)} \begin{bmatrix} K_1(\tau, \theta) & K_2(\tau, \theta) \end{bmatrix}$$

where

$$\begin{aligned} K_1(\tau, \theta) &= 76.930 - 1109.596\tau + 14.343\theta + 1569.878\tau^2 + 170.469\tau\theta - 9.158\theta^2 \\ K_2(\tau, \theta) &= 24.445 - 739.302\tau - 17.004\theta + 1136.874\tau^2 + 159.427\tau\theta + 3.174\theta^2 \\ \text{den}(\tau, \theta) &= -23.189 + 483.241\tau - 0.934\theta - 947.359\tau^2 + 3.140\tau\theta + 1.066\theta^2 \end{aligned}$$







**D-BSSE**  
Department of Biosystems  
Science and Engineering

# Concluding remarks



## Concluding statements

- Tractable conditions for analysis and control of LPV systems with piecewise constant parameters
- Extend quadratic and robust stability





## Concluding statements

- Tractable conditions for analysis and control of LPV systems with piecewise constant parameters
- Extend quadratic and robust stability

## Possible extensions

- Piecewise differentiable parameters (underway)
- Dynamic output feedback?
- Performance analysis, e.g.  $L_2$ -performance
- Nonlinear systems
- Homogeneous Lyapunov functions (non-conservative<sup>1</sup>)

ETH Zürich

<sup>1</sup> 

F. Wirth. A converse Lyapunov theorem for linear parameter-varying and linear switching systems, *SIAM Journal on Control and Optimization*,

2005

**D-BSSE**Department of Biosystems  
Science and Engineering

# Thank you for your attention