

**S-variables for the positivity check  
of matrix polynomials with matrix indeterminates**

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# Introduction

- Many mathematical tools in robust control for building LMI conditions
- Lyapunov, S-procedure, KYP, DG-scaling, IQC, Quadratic Separation, Finsler lemma, S-variables, Positivstellensatz, SOS, Polyá...
- ▲ Developed separately for specific uncertainties and view points
- ▲ Conservative SDP relaxations
- ▲ Hierarchies of relaxations with decreasing conservatism
- This presentation : Attempt to establish links between these tools
- ▲ Positivity of matrix polynomials with matrix indeterminates
- ▲ Continuation of work of S-variable approach [Ebihara]
- ▲ Strongly inspired by Quadratic Separation [Iwasaki]
- ▲ Connexions to be done with Generalized Frequency Variables [Hara]
- ▲ Technicalities linking SOS and S-variables [Sato]

# Motivation - Lyapunov

■ Stability of a linear system  $\dot{x} = Ax$

● All eigenvalues of  $A$  have negative real part

●  $sI - A$  is non singular for all  $s \in \overline{\mathbb{C}}_+$

●  $I - As^{-1}$  is non singular for all  $s^{-1} \in \overline{\mathbb{C}}_+$

●  $\exists \epsilon : (I - As^{-1})^*(I - As^{-1}) \succeq \epsilon I \succ 0$  for all  $s^{-1} + s^{-*} \geq 0$

▲ Matrix valued polynomial inequality constrained by a polynomial inequality

▲ Indeterminate is complex-valued  $s^{-1} \in \mathbb{C}$

●  $\exists P \succeq 0, \exists \epsilon > 0$  such that  $A^*P + PA \preceq -\epsilon I$

▲ **Equivalent** LMI formulation

▲  $P$  is a Lagrange-like multiplier such that  $P(s^{-1} + s^{-*}) \succeq 0$

## Motivation - Lyapunov & S-variables

■ Stability of a linear system  $\dot{x} = Ax$  with  $A \in CO\{A^{[1]}, \dots, A^{[\bar{v}]}\}$

$$CO\{A^{[1]}, \dots, A^{[\bar{v}]}\} = \left\{ A = \sum_{v=1}^{\bar{v}} \xi_v A^{[v]} : \xi_v \geq 0, \sum_{v=1}^{\bar{v}} \xi_v = 1 \right\}$$

●  $\lambda(A)$  have negative real part  $\forall A \in CO\{A^{[1]}, \dots, A^{[\bar{v}]}\}$

●  $I - As^{-1}$  is non singular  $\forall s^{-1} \in \overline{\mathbb{C}}_+, \forall A \in CO\{A^{[1]}, \dots, A^{[\bar{v}]}\}$

●  $\begin{bmatrix} I & -s^{-1}I \\ -A & I \end{bmatrix}$  is non singular  $\forall s^{-1} \in \overline{\mathbb{C}}_+, \forall A \in CO\{A^{[1]}, \dots, A^{[\bar{v}]}\}$

●  $\exists \epsilon : \begin{bmatrix} I & -s^{-1}I \\ -A & I \end{bmatrix}^* \begin{bmatrix} I & -s^{-1}I \\ -A & I \end{bmatrix} \succeq \epsilon I \succ 0$  for all  $s^{-1} + s^{-*} \geq 0$   
 $A \in CO\{A^{[1]}, \dots, A^{[\bar{v}]}\}$

▲ Matrix polynomial inequality with indeterminates constrained

by polynomial inequalities & polytopes

▲ Indeterminates are in independent rows and columns

## Motivation - Lyapunov & S-variables

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 $A \in CO\{A^{[1]}, \dots, A^{[\bar{v}]}\}$

▲ Matrix polynomial inequality with indeterminates constrained  
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●  $\exists S : \forall v = 1 \dots \bar{v}, \exists P^{[v]} \succeq 0$  such that

$$\begin{bmatrix} \epsilon I & P^{[v]} \\ P^{[v]} & \epsilon I \end{bmatrix} \preceq S \begin{bmatrix} A^{[v]} & -I \\ & \end{bmatrix} + (S \begin{bmatrix} A^{[v]} & -I \\ & \end{bmatrix})^*$$

▲ **Conservative** LMI formulation

▲  $P(A) = \sum_{v=1}^{\bar{v}} \xi_v P^{[v]}$ , parameter-dependent, s.t.  $P(A)(s^{-1} + s^{-*}) \succeq 0$

▲ S-variable copes with the polytopic uncertainty

# Motivation - DG-scalings

## Well-posedness of $\Delta \star M$

$\bullet \Delta = \begin{bmatrix} \delta_1 I_{r_1} & & 0 \\ & \ddots & \\ 0 & & \Delta_{\bar{k}} \end{bmatrix}$

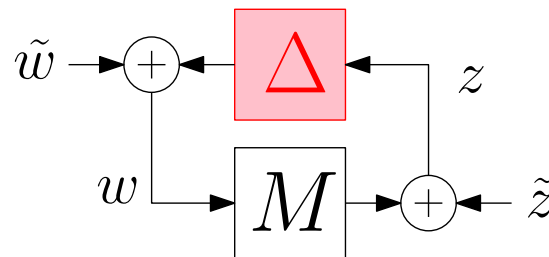
▼ independent uncertainties

▼ scalar repeated or matrix valued

▼ real or complex

▼ norm-bounded by 1 :  $|\delta_k| \leq 1$  or  $\|\Delta_k\| \leq 1$

$\bullet \star$  : feedback-loop



$\bullet M$  : Complex valued matrix

$\bullet$  Well-posedness : internal  $(w, z)$  bounded for all bounded disturbances  $(\tilde{w}, \tilde{z})$

# Motivation - DG-scalings

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▼ norm-bounded by 1 :  $|\delta_k| \leq 1$  or  $\|\Delta_k\| \leq 1$

$$\delta_k \in \mathbb{C}, \quad |\delta_k| \leq 1 \quad \Leftrightarrow \quad 1 \geq \delta_k^* \delta_k$$

$$\Delta_k \in \mathbb{C}^{m_{1k}, m_{2k}}, \quad \|\Delta_k\| \leq 1 \quad \Leftrightarrow \quad I \succeq \Delta_k^* \Delta_k$$

$$\delta_k \in \mathbb{R}, \quad |\delta_k| \leq 1 \quad \Leftrightarrow \quad -j\delta_k^* + j\delta_k = 0, \quad 1 \geq \delta_k^* \delta_k$$

$$\delta_k \in \mathbb{R}, \quad |\delta_k| \leq 1 \quad \Leftrightarrow \quad \delta_k \in CO\{-1, 1\}$$

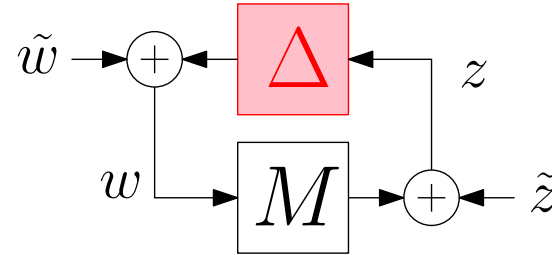
▲ Indeterminates constrained by polynomial inequalities/equalities & polytopes

▲ Uncertainties are repeated  $I_{r_k} \otimes \Delta_k$  (generalization of  $\delta_1 I_{r_1}$  to matrices)

# Motivation - DG-scalings

■ Well-posedness of  $\Delta \star M$

●  $\Delta = \dots$



●  $\star$  : feedback-loop

● Well-posedness : internal  $(w, z)$  bounded for all bounded disturbances  $(\tilde{w}, \tilde{z})$

$$\begin{bmatrix} I_{m_1} & \Delta \\ M & I_{m_2} \end{bmatrix} \begin{pmatrix} w \\ z \end{pmatrix} = \begin{pmatrix} \tilde{w} \\ \tilde{z} \end{pmatrix}$$

●  $\exists \epsilon > 0$  such that for all admissible  $\Delta$

$$(I_{m_2} - M\Delta)^*(I_{m_2} - M\Delta) \succeq \epsilon I_{m_2}$$

▲ Matrix polynomial inequality with indeterminates constrained

by polynomial inequalities/equalities & polytopes

▲ Indeterminates are in independent rows and columns ( $\Delta$  block-diagonal)



# Motivation - DG-scalings

## Well-posedness of $\Delta \star M$

$\bullet \Delta = \begin{bmatrix} I_{r_1} \otimes \Delta_1 & & 0 \\ & \ddots & \\ 0 & & I_{r_{\bar{k}}} \otimes \Delta_{\bar{k}} \end{bmatrix}$

- $\blacktriangledown$  polynomial inequality on  $\Delta_k$
- $\blacktriangledown$  polynomial equality on  $\Delta_k$
- $\blacktriangledown$  polytopic constraint  $\Delta_k$

$\bullet \exists \epsilon > 0$  such that for all admissible  $\Delta$

$$(I_{m_2} - M\Delta)^*(I_{m_2} - M\Delta) \succeq \epsilon I_{m_2}$$

## Conservative LMI condition

$$\exists D_k \succeq 0, G_k \quad : \quad \begin{bmatrix} I & M^* \end{bmatrix} \Theta(D_k, G_k) \begin{bmatrix} I \\ M \end{bmatrix} \succeq \epsilon I$$

$\blacktriangle \Theta(D_k, G_k)$  : linear in the decision variables

$\blacktriangle D_k, G_k$  : Lagrange-like multipliers w.r.t inequality and equality constraints

# Motivation - Proving positivity under constraints

- Robustness analysis of linear time-invariant systems
  
- Most problems can be recast as proving positivity of polynomials
  - ▼ matrix valued (semi-definite constraints)
  - ▼ indeterminates are matrices (or scalars), complex valued
  - ▼ constrained by polynomial inequalities, equalities and in polytopes
  
- Many LMI results in the literature,
  - ▼ in general **conservative** (problems are NP-hard)
  - ▲ some results are proved to be less conservative
  - ▲ on examples conservatism may vanish
  - ▲ duality of SDPs can extract worst case indeterminates (prove conservatism vanishes)
  - ▼ Numerical issues : limit size of LMIs using the structure of the data

# Introduction

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- ▲ Technicalities linking SOS and S-variables [Sato]

# Sum-Of-Squares

- [Lasserre], [Parillo], [Scherer], [Chesi] ... vast literature
  
- Goal : proving positivity of a matrix valued polynomial  $F(\delta) \succeq 0$ 
  - ▲ with scalar indeterminates  $\delta \in \mathbb{R}^{\bar{k}}$
  - ▲ constrained by scalar polynomial inequalities  $f_i(\delta) \geq 0$ .
  
- Some key methods for solving the problem using SDPs
  - ▲ Positivstellensatz
  - ▲ Polynomials modeled as quadratic functions of monomials
  - ▲ SDP relaxation
  - ▲ Hierarchies
  - ▲ Moment problem

# Sum-Of-Squares

■  $F(\delta) \succeq 0$  for all  $\delta \in \mathbb{R}^{\bar{k}}$  such that  $f_i(\delta) \geq 0$

● Positivstellensatz [Putinar] : Polynomial sum-of-squares multipliers for each constraint

$$\exists D_i(\delta) \text{ SOS} \quad : \quad F_D(\delta) = d_0(\delta)F(\delta) + \sum D_i(\delta)f_i(\delta) \text{ SOS}$$

▲ Lossless when taking polynomials  $D_i(\delta)$  with sufficiently high order

▲ Variants of Positivstellensatz are : S-procedure, D-scalings, Lagrange multipliers...

# Sum-Of-Squares

■  $F(\delta) \succeq 0$  for all  $\delta \in \mathbb{R}^{\bar{k}}$  such that  $f_i(\delta) \geq 0$

● Positivstellensatz : Polynomial positive multipliers for each constraint

$$\exists D_i(\delta) \text{ SOS} \quad : \quad F_D(\delta) = d_0(\delta)F(\delta) + \sum D_i(\delta)f_i(\delta) \text{ SOS}$$

● Express the polynomials as a quadratic functions of monomials

$$\delta^{\{0:p\}*} = \left[ \begin{array}{cccccccc} I & \delta_1 I & \delta_2 I & \dots & \delta_1 \delta_2 I & \dots & \delta_1^{p_1} \delta_2^{p_2} I & \dots \end{array} \right]$$

$$D_i(\delta) = \delta^{\{0:p\}*} Q_i(d_i) \delta^{\{0:p\}}, \quad F_D(\delta) = \delta^{\{0:p\}*} Q(d) \delta^{\{0:p\}}$$

$Q_i(d_i)$  and  $Q(d)$  are linear in decision variables  $d_i$  (coef. of polynomials  $D_i(\delta)$ )

▲ Quadratic representations are not unique

$$F_D(\delta) = \delta^{\{0:p\}*} (Q(d) + V) \delta^{\{0:p\}} \quad \text{whatever } V \text{ s.t. } \delta^{\{0:p\}*} V \delta^{\{0:p\}} = 0 \quad \forall \delta$$

these are linear constraints on  $V$  denoted  $V \in \mathcal{N}(\delta^{\{0:p\}})$

# Sum-Of-Squares

■  $F(\delta) \succeq 0$  for all  $\delta \in \mathbb{R}^{\bar{k}}$  such that  $f_i(\delta) \geq 0$

● Positivstellensatz + monomials (for large enough order  $p$ )

$$\exists \delta^{\{0:p\}*} (Q_i(d_i)) \delta^{\{0:p\}} \text{ SOS} \quad : \quad \delta^{\{0:p\}*} (Q(d)) \delta^{\{0:p\}} \text{ SOS}$$

● SDP relaxation (for fixed order  $p$ )

$$\exists V_i \in \mathcal{N}(\delta^{\{0:p\}}) \quad : \quad Q_i(d_i) + V_i \succeq 0 \quad , \quad Q(d) + V \succeq 0$$

▲ LMI : convex and exist efficient solvers (for not too large size problems)

▼ Conservative

▼ Parameterization of  $\mathcal{N}(\delta^{\{0:p\}})$  is not easy

# Sum-Of-Squares

■  $F(\delta) \succeq 0$  for all  $\delta \in \mathbb{R}^{\bar{k}}$  such that  $f_i(\delta) \geq 0$

● Positivstellensatz + monomials + SDP relaxation (for fixed order  $p$ )

$$\exists V_i \in \mathcal{N}(\delta^{\{0:p\}}) \quad : \quad Q_i(d_i) + V_i \succeq 0 \quad , \quad Q(d) + V \succeq 0$$

● Hierarchies : increasing  $p$ , the order of the polynomials (and scalings)

▲ Conservatism decreases (rapidly)

▲ Under mild assumption conservatism vanishes (for finite and low orders)

▼ Numerical burden increases rapidly (need to exploit the structure)

● Dual of the LMIs is the relaxation of generalized moment problem



# Motivation - Proving positivity under constraints

- Robustness analysis of linear time-invariant systems
- Most problems can be recast as proving positivity of polynomials
  - ▼ matrix valued (semi-definite constraints)
  - ▼ indeterminates are matrices (or scalars), complex valued
  - ▼ constrained by polynomial inequalities, equalities and in polytopes
- Many LMI results in the literature,
  - ▼ in general **conservative** (problems are NP-hard)
  - ▲ some results are proved to be less conservative
  - ▲ on examples conservatism may vanish
  - ▲ duality of SDPs can extract worst case indeterminates (prove conservatism vanishes)
  - ▼ Numerical issues : limit size of LMIs using the structure of the data
- Same characteristics in the SOS-Moment framework
  - ▼ Unification needs to manipulate polynomials with matrix indeterminates

# Polynomials with matrix indeterminates

■ Goal : proving  $F(\Delta) = F(I_{r_1} \otimes \Delta_1, \dots, I_{r_{\bar{k}}} \otimes \Delta_{\bar{k}}) \succeq 0$

▼ under polynomials + polytopic constraints of the type

$$F_{ik}(\Delta_k) \succeq 0, \quad F_{ek}(\Delta_k) = 0, \quad \Delta_k \in CO\{\Delta_k^{[1]}, \dots, \Delta_k^{[\bar{v}_k]}\}$$

# Polynomials with matrix indeterminates

- Monomials with matrix indeterminates  $\Delta_k \in \mathbb{C}^{m_{1k} \times m_{2k}}$

$$\begin{aligned} \Delta_k \{0\} &= I_{m_{2k}} & \Delta_k \{1\} &= \Delta_k, & \Delta_k \{2\} &= \Delta_k^* \Delta_k, \\ \Delta_k \{3\} &= \Delta_k \Delta_k^* \Delta_k, & \Delta_k \{4\} &= \Delta_k^* \Delta_k \Delta_k^* \Delta_k, & \dots & \end{aligned}$$

- ▲ Matrix with monomials from degree 0 to degree  $p_k$  :

$$\begin{aligned} \Delta_k \{0:p_k\} &= \begin{bmatrix} \Delta_k \{0\} \\ \vdots \\ \Delta_k \{p_k\} \end{bmatrix} = \begin{bmatrix} I_{m_{2k}} \\ \Delta_k \\ \Delta_k^* \Delta_k \\ \vdots \end{bmatrix}, & (I_{r_k} \otimes \Delta_k) \{0:p_k\} &= \begin{bmatrix} I_{r_k} \otimes \Delta_k \{0\} \\ \vdots \\ I_{r_k} \otimes \Delta_k \{p_k\} \end{bmatrix} \\ \\ \Delta \{0:p\} &= \begin{bmatrix} (I_{r_1} \otimes \Delta_1) \{0:p_1\} & & 0 \\ & \ddots & \\ 0 & & (I_{r_{\bar{k}}} \otimes \Delta_{\bar{k}}) \{0:p_{\bar{k}}\} \end{bmatrix} \end{aligned}$$

# Polynomials with matrix indeterminates

- With assumption that  $\Delta_k$  enter in independent columns

$$F(\Delta) = \Delta^{\{0:p\}*} (F_0 + F_1^* F_1) \Delta^{\{0:p\}} \succeq 0$$

under constraints

$$F_{ik}(\Delta_k) = \Delta_k^{\{0:p_k\}*} \Phi_{ik} \Delta_k^{\{0:p_k\}} \succeq 0, \quad \Delta_k \in CO\{\Delta_k^{[1]}, \dots, \Delta_k^{[\bar{v}_k]}\}$$

$$F_{ek}(\Delta_k) = \Delta_k^{\{0:p_k\}*} \Phi_{ek} \Delta_k^{\{0:p_k\}} = 0,$$

- "Positivstellensatz" : Exist  $D_k(\Delta) \succeq 0, G_k(\Delta) = G_k^*(\Delta)$  such that

$$\Delta^{\{0:p\}*} (F_0 + F_1^* F_1 + \text{diag}(\dots D_k(\Delta) \boxtimes \Phi_{ik} + G_k(\Delta) \boxtimes \Phi_{ek} \dots)) \Delta^{\{0:p\}} \succeq 0$$

under polytopic constraints  $\Delta_k \in CO\{\Delta_k^{[1]}, \dots, \Delta_k^{[\bar{v}_k]}\}$  ( $\bar{v}_k = 0$  if no constraint)

- ▼ Conservative ?
- ▼ How to build the SDP relaxation for fixed order  $p$  ?

# S-variables and SDPs for SOS

- Exists affine  $H_k(\Delta_k) = J_0 + J_1(I \otimes \Delta_k)J_2 + J_3(I \otimes \Delta_k^*)J_4$   
such that  $\Delta_k^{\{0:p_k\}}$  spans the null space of  $H_k(\Delta_k)$

$$\underbrace{\begin{bmatrix} \Delta_k & -I & 0 & 0 \\ 0 & \Delta_k^* & -I & 0 \\ 0 & 0 & \Delta_k & -I \end{bmatrix}}_{M_k(\Delta_k)} \underbrace{\begin{bmatrix} I \\ \Delta_k \\ \Delta_k^* \Delta_k \\ \Delta_k \Delta_k^* \Delta_k \end{bmatrix}}_{\Delta_k^{\{0:p_k\}}} = 0$$

# S-variables and SDPs for SOS

- $\Delta_k^{\{0:p_k\}}$  null space of  $H_k(\Delta_k) = J_0 + J_1(I \otimes \Delta_k)J_2 + J_3(I \otimes \Delta_k^*)J_4$

- SDP relaxation for  $\Delta_k$  in polytope

assuming  $\Psi(X(\Delta))$  is affine of indeterminate-dependent  $X(\Delta)$

$$\exists S_k, X^{[v_k]} : \Psi(X^{[v_k]}) + S_k H_k(\Delta_k^{[v_k]}) + (S_k H_k(\Delta_k^{[v_k]}))^* \succeq 0 \quad \forall v_k = 1 \dots \bar{v}_k$$

⇓

$$\exists X(\Delta) : \Delta_k^{\{0:p_k\}*} \Psi(X(\Delta)) \Delta_k^{\{0:p_k\}} \succeq 0 \quad \forall \Delta_k \in CO\{\Delta_k^{[1]}, \dots, \Delta_k^{[\bar{v}_k]}\}$$

# S-variables and SDPs for SOS

## ● Proof

▲ Affine  $\Psi$  and  $M_k$  and  $\Delta_k = \sum_{v_k=1}^{\bar{v}_k} \xi_v \Delta^{[v_k]}$

$$\exists S_k, X^{[v_k]} : \Psi(X^{[v_k]}) + S_k H_k(\Delta_k^{[v_k]}) + (S_k H_k(\Delta_k^{[v_k]}))^* \succeq 0 \quad \forall v_k = 1 \dots \bar{v}_k$$



$$\exists S_k, : \Psi(X(\Delta)) + S_k H_k(\Delta_k) + (S_k H_k(\Delta_k))^* \succeq 0 \quad \forall \xi_v \geq 0, \quad \sum_{v=1}^{\bar{v}} \xi_v = 1$$

$$X(\Delta) = \sum_{v_k=1}^{\bar{v}_k} \xi_v X^{[v_k]}$$

▲ By congruence with the fact that  $H_k(\Delta_k) \Delta_k^{\{0:p_k\}} = 0$



$$\exists X(\Delta) : \Delta_k^{\{0:p_k\}*} \Psi(X(\Delta)) \Delta_k^{\{0:p_k\}} \succeq 0 \quad \forall \Delta_k \in CO\{\Delta_k^{[1]}, \dots, \Delta_k^{[\bar{v}_k]}\}$$

▼ Conservatism comes for the choice of indeterminate-independent  $S_k$

## S-variables and SDPs for SOS

●  $\Delta_k^{\{0:p_k\}}$  null space of  $H_k(\Delta_k) = J_0 + J_1(I \otimes \Delta_k)J_2 + J_3(I \otimes \Delta_k^*)J_4$

● SDP relaxation for unbounded  $\Delta_k$

$$\exists S_k, X : \Psi(X) + S_k H_k(0) + (S_k H_k(0))^* \succeq 0 \quad \forall v_k = 1 \dots \bar{v}_k$$

$$S_k = \begin{bmatrix} J_4^T & J_2^T \end{bmatrix} \begin{bmatrix} T_k \otimes I & 0 \\ 0 & T_k \otimes I \end{bmatrix} \begin{bmatrix} J_1^T \\ J_3^T \end{bmatrix}, \quad T_k = -T_k^*$$

⇓

$$\exists X : \Delta_k^{\{0:p_k\}*} \Psi(X) \Delta_k^{\{0:p_k\}} \succeq 0 \quad \forall \Delta_k$$

▲ Structured S-variables provides a parameterization of  $\mathcal{N}(\Delta_k^{\{0:p_k\}})$

▼ Is  $\mathcal{N}(\Delta_k^{\{0:p_k\}})$  fully parameterized in this way?

▲ SOS-Moment SDP relaxations are a special case of S-variable relaxation



# S-variables and SDPs for SOS

■ Main result (combining described techniques)

$$\hat{F}_1^{\perp*} \left( \hat{F}_0 - \text{diag} \left( \begin{array}{c} \vdots \\ \hat{D}_k^{[v_{\mathcal{K}_2}]} \boxtimes \Phi_{ik} + \hat{G}_k^{[v_{\mathcal{K}_2}]} \boxtimes \Psi_{ek} \\ \vdots \end{array} \right) \right) \hat{F}_1^{\perp} + \sum_{k \in \mathcal{K}_1} \left\{ \hat{F}_1^{\perp*} \hat{S}_k^{[v_{\mathcal{K}_2}]} \hat{H}_k(0) \hat{F}_1^{\perp} \right\}^{\mathcal{H}} + \sum_{k \in \mathcal{K}_2} \left\{ \tilde{S}_k^{[v_{\mathcal{K}_2} \setminus k]} \hat{H}_k(\Delta_k^{[v_k]}) \hat{F}_1^{\perp} \right\}^{\mathcal{H}} \succeq 0$$

▲  $\mathcal{K}_1, \mathcal{K}_2$  : uncertainties without and with polytopic constraints respectively

▼ Large number of constraints and decision variables (when several polytopes)

▼ Are all the decision variables needed ?

▼ Can we build hierarchies as in usual SOS methods ?

▲ When applied to special cases we get exactly the same LMI conditions

# Conclusion

- ▲ Ongoing work to make links between many existing results in Robust Control
- ▲ Used tool is inspired by SOS-Moments relaxations
- ▲ Motivation for dealing with polynomial matrix inequalities of matrix indeterminates
- ▼ Sub-case of all possible polynomial matrix inequalities of matrix indeterminates
- ▼ At this stage does not allow to build non-existing results
- ▼ Numerical experiments to be done