

About S-Variables and Descriptor Systems Dimitri Peaucelle

▶ To cite this version:

Dimitri Peaucelle. About S-Variables and Descriptor Systems. Workshop on Automatic Control Theory Japan-France Research Networking, Nov 2024, Kyoto (Japan), Japan. hal-04926040

HAL Id: hal-04926040 https://hal.science/hal-04926040v1

Submitted on 3 Feb 2025

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About S-Variables and Descriptor Systems

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Workshop on Automatic Control Theory Japan-France Research Networking Kyoto - 8 November 2024







Abstract



• [Aizerman and Gantmakher, 1964], Lossless [Yakubovitch, 1971]

$$\eta^* Q \eta < 0 \quad orall \eta^* Q_1 \eta \leq 0 \quad \Leftrightarrow \quad \exists s > 0 \; : \; Q \prec s Q_1$$

▲ Extensions: bounded real lemma, passivity, dissipation, KYP

Lossy

$$\eta^* Q\eta < 0 \quad orall \left\{ egin{array}{cc} \eta^* Q_1 \eta \leq 0 \ \eta^* Q_2 \eta \leq 0 \end{array}
ight. \quad \Leftarrow \quad \exists s_1 > 0, s_2 > 0 \ : \ Q \prec s_1 Q_1 + s_2 Q_1 \end{array}
ight.$$

 \checkmark Extensions such as μ -analysis, IQCs, quadratic separation, are (in most cases) conservative



• S-procedure

$\eta^* Q \eta < 0 \quad orall \eta^* Q_1 \eta \leq 0 \quad \Leftrightarrow \quad \exists s > 0 \; : \; Q \prec s Q_1$

S-Variable approach

$$\eta^* Q \eta < 0 \quad \forall M \eta = 0 \quad \Leftrightarrow \quad M^{\perp T} Q M^{\perp} \prec 0 \quad \Leftrightarrow \quad \exists S : Q \prec S M + M^* S^* = \{SM\}^{\mathcal{H}}$$

S-procedure based proof [Ebihara et al., 2015]: $Q_1 = M^*M$ and $S = s\frac{1}{2}M^*$ Many new degrees of freedom

Increased size of LMIs & many more decision variables



Lossless

$$\eta^* Q \eta < 0 \quad \forall M \eta = 0 \quad \Leftrightarrow \quad \exists S : Q \prec \{SM\}^{\mathcal{H}}$$

• Lossy but Robust [Peaucelle et al., 2000]

$$\{ \eta^* Q(\zeta)\eta < 0 \quad \forall M(\zeta)\eta = 0 \} \quad \forall \zeta \in [0 \ 1] \quad \Leftarrow \quad \exists S : \begin{cases} Q^{[0]} \prec \{SM^{[0]}\}^{\mathcal{H}} \\ Q^{[1]} \prec \{SM^{[1]}\}^{\mathcal{H}} \end{cases}$$

$$Q({oldsymbol{\zeta}})=(1-{oldsymbol{\zeta}})Q^{[0]}+{oldsymbol{\zeta}}Q^{[1]}$$
 , $M({oldsymbol{\zeta}})=(1-{oldsymbol{\zeta}})M^{[0]}+{oldsymbol{\zeta}}M^{[1]}$

• Lossy but good for design (see WUDS 2023 Kyoto [Peaucelle, 2023])

$$\begin{cases} \eta^* Q \eta < 0 \quad \forall M_1 \eta = 0 \\ \eta^* Q \eta < 0 \quad \forall M_2 \eta = 0 \end{cases} \quad \Leftarrow \quad \exists S = M_2^* V : Q \prec \{SM_1\}^{\mathcal{H}}$$

Heuristics for Control (and Robust Control) Design

 \checkmark Convex only for chosen M_2



Lossless

$$\eta^* Q \eta < 0 \quad \forall M \eta = 0 \quad \Leftrightarrow \quad \exists S : Q \prec \{SM\}^{\mathcal{H}}$$

• Lossy but Robust [Peaucelle et al., 2000]

$$\{ \eta^* Q(\zeta)\eta < 0 \quad \forall M(\zeta)\eta = 0 \} \quad \forall \zeta \in [0 \ 1] \quad \Leftarrow \quad \exists S : \left\{ \begin{array}{c} Q^{[0]} \prec \{SM^{[0]}\}^{\mathcal{H}} \\ Q^{[1]} \prec \{SM^{[1]}\}^{\mathcal{H}} \end{array} \right. \\ \left[\begin{array}{c} Q(\zeta) \\ M(\zeta) \end{array} \right] = (1 - \zeta) \left[\begin{array}{c} Q^{[0]} \\ M^{[0]} \end{array} \right] + \zeta \left[\begin{array}{c} Q^{[1]} \\ M^{[1]} \end{array} \right] \in \operatorname{CO} \left\{ \left[\begin{array}{c} Q^{[0]} \\ M^{[0]} \end{array} \right], \left[\begin{array}{c} Q^{[1]} \\ M^{[1]} \end{array} \right] \right\}$$

Efficient for analysis of parametric uncertain systems, including polynomial and rational

Example for Robustness analysis

$$V(x,\zeta) = x^* P(\zeta) x > 0 \quad \dot{V}(x,\zeta) < 0 \quad \forall \ddot{y} + \alpha_1(\zeta) \dot{y} + \frac{1}{\alpha_2(\zeta)} y = 0 \qquad \alpha_i(\zeta) \text{ affine of } \zeta \in \mathcal{C}$$

Affine descriptor version (alternative to building LFTs A smaller size)

$$\dot{V}(\mathbf{x}, \boldsymbol{\zeta}) < \mathbf{0} \qquad orall \left\{ egin{array}{ll} \ddot{\mathbf{y}} + lpha_{1}(\boldsymbol{\zeta})\dot{\mathbf{y}} + \pi = \mathbf{0} \ lpha_{2}(\boldsymbol{\zeta})\pi = \mathbf{y} \end{array}
ight.$$

State-space descriptor version with $\mathbf{x} = \begin{pmatrix} \mathbf{y} & \dot{\mathbf{y}} \end{pmatrix}^T$

$$\eta^{*} \underbrace{ \begin{bmatrix} 0 & 0 & P(\zeta) \\ 0 & 0 & 0 \\ P(\zeta) & 0 & 0 \end{bmatrix}}_{Q(P(\zeta))} \eta < 0 \qquad \forall \underbrace{ \begin{bmatrix} 1 & 0 & 0 & 0 & -1 \\ 0 & 1 & 1 & 0 & \alpha_{1}(\zeta) \\ 0 & 0 & \alpha_{2}(\zeta) & -1 & 0 \end{bmatrix}}_{M(\zeta)} \underbrace{ \begin{pmatrix} \dot{x} \\ \pi \\ x \end{pmatrix}}_{\eta} = 0$$

▲ (Lossy) S-variable result: Robust stability holds true if $\exists P^{[0]} \succ 0$, $P^{[1]} \succ 0$ and S: $Q(P^{[0]}) \prec \{SM^{[0]}\}^{\mathcal{H}}$, $Q(P^{[0]}) \prec \{SM^{[1]}\}^{\mathcal{H}}$

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Lifting by taking higher order derivatives:

$$\begin{bmatrix} 1\\ \frac{d}{dt} \end{bmatrix} (y^{(2)} + \alpha_1(\zeta)y^{(1)} + \frac{1}{\alpha_2(\zeta)}y) = \begin{bmatrix} y^{(2)} + \alpha_1(\zeta)y^{(1)} + \frac{1}{\alpha_2(\zeta)}y \\ y^{(3)} + \alpha_1(\zeta)y^{(2)} + \frac{1}{\alpha_2(\zeta)}y^{(1)} \end{bmatrix} = 0$$

State-space descriptor version with $\hat{x}_1 = \begin{pmatrix} y & y^{(1)} & y^{(2)} \end{pmatrix}'$

$$\begin{bmatrix} 1 & 0 & 0 & 0 & 0 & 0 & -1 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 & 0 & -1 \\ 0 & 1 & 0 & 1 & 0 & 0 & \alpha_1(\zeta) & 0 \\ 0 & 0 & 1 & 0 & 1 & 0 & 0 & \alpha_1(\zeta) \\ 0 & 0 & 0 & \alpha_2(\zeta) & 0 & -1 & 0 \end{bmatrix} \begin{pmatrix} \dot{\chi}_1 \\ \hat{\pi}_1 \\ \hat{\chi}_1 \end{pmatrix} = 0$$



Lifting by causal / non causal, rationnally parameter-dependent, filters :

$$\begin{bmatrix} 1\\ H(s,\zeta) \end{bmatrix} (y^{(2)} + \alpha_1(\zeta)y^{(1)} + \frac{1}{\alpha_2(\zeta)}y) = 0$$

- ▲ (Lossy) S-variable result applies and gives a less conservative, larger size, LMI
- ▲ Stable causal filters can be chosen with zero initial conditions. Link with recent work by Scherer?
- Connexion to dynamic IQC multipliers?
- Vertice of filters?
- Parameter-dependent lifting and SOS type hierarchies?
- Hierarchies of LMI? Convergence to exactness?

The S-variable Robustness conditions for $\{ \eta^* Q(\zeta)\eta < 0 \quad \forall M(\zeta)\eta = 0 \} \quad \forall \zeta \in [0 \ 1]$ involve LMI constraints on vertices

$$Q(\mathcal{P}^{[0]}) \prec \{\mathcal{SM}^{[0]}\}^{\mathcal{H}} \quad , \quad Q(\mathcal{P}^{[1]}) \prec \{\mathcal{SM}^{[1]}\}^{\mathcal{H}}$$

For each constraint exist a dual variable $H^{[0]} \succeq 0$ and $H^{[1]} \succeq 0$ respectively.

▲ [Ebihara et al., 2015] If the primal LMIs are unfeasible (or marginally feasible) and dual variables satisfy

 $\exists \zeta \in [0 \ 1]$: $(1 - \zeta) H^{[0]} + \zeta H^{[1]} = 0$

then ζ is a worst-case such that $\exists M(\zeta)\eta : \eta^*Q(\zeta)\eta \geq 0.$

Worst-case extraction

▲ In practice, solve convex 2nd order cone problem $\zeta^* = \min_{\zeta} \|(1-\zeta)H^{[0]} + \zeta H^{[1]}\|_2$. Gives rather good close to "worst case" values.

• One polytopic uncertainty

$$\eta^* Q(P(\zeta))\eta < 0 \qquad \forall M(\zeta)\eta = 0 \qquad M(\zeta) = \sum_{\nu=1}^{\bar{\nu}} \zeta_{\nu} M^{[\nu]} \quad \zeta_{\nu} \ge 0 \qquad \sum_{\nu=1}^{\bar{\nu}} \zeta_{\nu} = 1$$

Robustness analysis w.r.t polytopic uncertainties

▲ Holds true if $\exists P^{[v=1...\bar{v}]} \succ 0$ and *S*:

$$Q(P^{[v]}) \prec \{SM^{[v]}\}^{\mathcal{H}} \quad \forall v = 1 \dots \bar{v}.$$

Proof: By convexity, the LMIs imply

$$Q(P(\zeta)) \prec \{SM(\zeta)\}^{\mathcal{H}} \quad orall \zeta_{m{v}} \geq 0 \quad \sum_{m{v}=1}^{ar{v}} \zeta_{m{v}} = 1$$

where $P(\zeta) = \sum_{\nu=1}^{\bar{\nu}} \zeta_{\nu} P^{[\nu]}$, and concludes the proof using the S-variable result.

• Two (or more) polytopic uncertainties [Is this new?]

$$\eta^* Q(P(\zeta_1,\zeta_2))\eta < 0 \qquad \forall \left[\begin{array}{c} M_1(\zeta_1) \\ M_2(\zeta_2) \end{array} \right] \eta = 0 \qquad M_i(\zeta_i) = \sum_1^{\bar{v}_i} \zeta_{i,v_i} M_i^{[v_i]} \quad \zeta_{i,v_i} \ge 0 \quad \sum_{v_1=1}^{\bar{v}_i} \zeta_{i,v_i} = 1$$

Robustness analysis w.r.t polytopic uncertainties

▲ Holds true if $\exists P^{[v_1=1...\bar{v}_1,v_2=1...\bar{v}_2]} \succ 0$, $S_1^{v_2=1...\bar{v}_2}$ and $S_2^{v_1=1...\bar{v}_1}$:

$$Q(P^{[v_1,v_2]}) \prec \left\{ S_1^{[v_2]} M_1^{[v_1]} + S_2^{[v_1]} M_2^{[v_2]} \right\}^{\mathcal{H}} \quad \forall v_i = 1 \dots \bar{v}_i$$

Proof: By convexity, the LMIs imply

$$Q(P(\zeta_1,\zeta_2)) \prec \{S_1(\zeta_2)M_1(\zeta_1) + S_2(\zeta_1)M_2(\zeta_2)\}^{\mathcal{H}} = \left\{ \begin{bmatrix} S_1(\zeta_2) & S_2(\zeta_1) \end{bmatrix} \begin{bmatrix} M_1(\zeta_1) \\ M_2(\zeta_2) \end{bmatrix} \right\}^{\mathcal{H}}$$

where $P(\zeta_1, \zeta_2) = \sum_{\nu_1=1}^{\bar{\nu}_1} \sum_{\nu_2=1}^{\bar{\nu}_2} \zeta_{1,\nu_1} \zeta_{2,\nu_2} P^{[\nu_1,\nu_2]}$, $S_i(\zeta_k) = \sum_{\nu_k=1}^{\bar{\nu}_k} \zeta_{k,\nu_k} S_i^{[\nu_k]}$ and concludes the proof using the S-variable result.

Example for Robustness analysis

At the expense of increasing the size of π , and reordering of rows, the $M(\zeta) = \begin{bmatrix} M_1(\zeta_1) \\ M_2(\zeta_2) \end{bmatrix}$ is always achievable.

 $T(\zeta) = N^*(\zeta)RN(\zeta) \prec 0$ $N(\zeta)$ is polynomial (or rational) of $\{\zeta_1, \dots, \zeta_{\bar{k}}\}$

Example for negativity of matrix polynomials

Equivalently written as

 $w^* R w < 0$ $\forall w = N(\zeta) z$

Descriptor version

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$$\begin{pmatrix} w \\ z \end{pmatrix}^* \begin{bmatrix} R & 0 \\ 0 & 0 \end{bmatrix} \begin{pmatrix} w \\ z \end{pmatrix} < 0 \qquad \forall \begin{bmatrix} -I & N(\zeta) \end{bmatrix} \begin{pmatrix} w \\ z \end{pmatrix} = 0$$

At the expense of including additional algebraically constrained vectors π

$$\eta^* \begin{bmatrix} R & 0 \\ 0 & 0 \end{bmatrix} \eta = \eta^* Q \eta < 0 \qquad \forall \begin{bmatrix} M_1(\zeta_1) \\ \vdots \\ M_{\bar{k}}(\zeta_{\bar{k}}) \end{bmatrix} \eta = 0 \qquad M_k(\zeta_k) \text{ affine of } \zeta_k$$

$$N(\zeta) = \begin{bmatrix} (1+2\zeta_1)(4+\zeta_2) & (1+2\zeta_1)\zeta_2\\ (4+\zeta_2) & \zeta_1+\zeta_2 \end{bmatrix}$$

Example for building affine descriptor forms

Decomposed in canonic monomial basis [Peaucelle and Sato, 2009]

$$N(\zeta)z = \begin{bmatrix} 4+8\zeta_1 & 0\\ 4 & \zeta_1 \end{bmatrix} z + \begin{bmatrix} 1+2\zeta_1 & 1+2\zeta_1\\ 1 & 1 \end{bmatrix} \zeta_2 z$$

gives a systematic way of building affine descriptor form for $w = N(\zeta)z$

$$\begin{bmatrix} -1 & 0 & | & 1+2\zeta_1 & 1+2\zeta_1 & | & 4+8\zeta_1 & 0 \\ 0 & -1 & 1 & 1 & | & 4 & \zeta_1 \\ 0 & 0 & | & -1 & 0 & | & \zeta_2 & 0 \\ 0 & 0 & 0 & -1 & | & 0 & \zeta_2 \end{bmatrix} \begin{pmatrix} w \\ \pi \\ z \end{pmatrix} = 0$$

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$N(\zeta) = \begin{bmatrix} (1+2\zeta_1)(4+\zeta_2) & (1+2\zeta_1)\zeta_2 \\ (4+\zeta_2) & \zeta_1+\zeta_2 \end{bmatrix}$

Example for building affine descriptor forms

Use Linear-Fractional Transformation (eg. in Robust Control Toolbox or LFR toolbox)

 $N(\zeta) = N_{11} + N_{12}\Delta(I - N_{22}\Delta)^{-1}N_{21}$ Δ diagonal matrix of ζ_1, ζ_2

gives an easy (but usually large) descriptor form for $w = N(\zeta)z$

$$\left[\begin{array}{ccc} -I & N_{11} & N_{12}\Delta \\ 0 & N_{12} & I - N_{22}\Delta \end{array}\right]\eta = 0$$

at the expense or reordering the rows the uncertainties appear as needed.

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LAAS Example for building affine descriptor forms

$$N(\zeta) = \begin{bmatrix} (1+2\zeta_1)(4+\zeta_2) & (1+2\zeta_1)\zeta_2\\ (4+\zeta_2) & \zeta_1+\zeta_2 \end{bmatrix}$$

▲ No need to decompose in canonic monomial basis, by exploiting the structure:

$$\begin{bmatrix} -1 & 0 & | & 1+2\zeta_1 & | & 0 & 0 \\ 0 & -1 & | & 1 & | & 0 & \zeta_1 \\ 0 & 0 & | & -1 & | & 4+\zeta_2 & \zeta_2 \end{bmatrix} \begin{pmatrix} w \\ \hat{\pi} \\ z \end{pmatrix} = 0$$

 $\{ \eta^* Q(\zeta)\eta < 0 \quad \forall M(\zeta)\eta = 0 \} \quad \forall \zeta \in [0 \ 1] \quad \Leftarrow \quad \exists S : \ Q(P^{[v]}) \prec \{SM^{[v]}\}^{\mathcal{H}}$

S-Variable - reducing the computational burden

\vee S is of the size of M^* . May become very large.

Veed for efficient software to build limited size $M(\zeta)$, project to do so in Julia with Ch. Louembet Veed to find systematic reduction (exact / approximate) tools for affine descriptor models

▲ Non conservative size reduction of LMIs if $M(\zeta) = T \begin{bmatrix} \hat{M}(\zeta) \\ \check{M} \end{bmatrix}$, T invertible

$$\exists \hat{S} : \begin{cases} \check{M}^{\perp T} Q_0 \check{M}^{\perp} \prec \{\hat{S} \hat{M}_0 \check{M}^{\perp}\}^{\mathcal{P}} \\ \check{M}^{\perp T} Q_1 \check{M}^{\perp} \prec \{\hat{S} \hat{M}_1 \check{M}^{\perp}\} \end{cases}$$

Exists also a (conservative but efficient) size reduction of LMIs

if
$$M(\zeta) = \begin{bmatrix} \hat{M}(\zeta) & \breve{M} \end{bmatrix} T$$
 and $\begin{bmatrix} 0 \\ I \end{bmatrix}^* T^{-*}Q(\zeta)T^{-1} \begin{bmatrix} 0 \\ I \end{bmatrix} = \hat{Q}_{22} \succeq 0.$

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S-Variable - case of unbounded uncertainties

Bounded uncertainties, where $Q(\zeta) = (1-\zeta)Q^{[0]} + \zeta Q^{[1]}$, $M(\zeta) = (1-\zeta)M^{[0]} + \zeta M^{[1]}$

$$\{ \eta^* Q(\boldsymbol{\zeta})\eta < 0 \quad \forall M(\boldsymbol{\zeta})\eta = 0 \} \quad \forall \boldsymbol{\zeta} \in [0 \ 1] \quad \Leftarrow \quad \exists S : \begin{cases} Q^{[0]} \prec \{SM^{[0]}\}^{\mathcal{H}} \\ Q^{[1]} \prec \{SM^{[1]}\}^{\mathcal{H}} \end{cases}$$

Unbounded uncertainties [Peaucelle and Sato, 2009] where $M(\zeta) = M^{[0]} + \zeta M^{[\infty]}$

$$\{ \eta^* Q \eta < 0 \quad \forall M(\zeta) \eta = 0 \} \quad \forall \zeta \quad \Leftarrow \quad \exists G = -G^* : Q \prec \{ M^{[\infty]*} G M^{[0]} \}^{\mathcal{H}}$$

▲ Surprisingly simple proof

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$$Q \prec \{M^{[\infty]*}GM^{[0]}\}^{\mathcal{H}} = \{M^{[\infty]*}GM(\zeta)\}^{\mathcal{H}} - \zeta\{M^{[\infty]*}GM^{[\infty]}\}^{\mathcal{H}}$$

last term is zero $M^{[\infty]*}GM^{[\infty]}$ $\mathcal{H} = M^{[\infty]*}(G + G^*)M^{[\infty]} = 0$. Hence

$$Q \prec \{SM(\zeta)\}^{\mathcal{H}}$$
, with $S = M^{[\infty]*}G$

LAAS S-Variable - case of unbounded uncertainties CNRS Bounded uncertainties, where $Q(\zeta) = (1 - \zeta)Q^{[0]} + \zeta Q^{[1]}$, $M(\zeta) = (1 - \zeta)M^{[0]} + \zeta M^{[1]}$ $\{ \eta^* Q(\zeta)\eta < 0 \quad \forall M(\zeta)\eta = 0 \} \quad \forall \zeta \in [0 \ 1] \quad \Leftarrow \quad \exists S : \begin{cases} Q^{[0]} \prec \{SM^{[0]}\}^{\mathcal{H}} \\ Q^{[1]} \prec \{SM^{[1]}\}^{\mathcal{H}} \end{cases}$ $M(\boldsymbol{\zeta}) = M^{[0]} + \boldsymbol{\zeta} M^{[\infty]}$ Unbounded uncertainties [Peaucelle and Sato, 2009] where $\{ \eta^* Q\eta < 0 \quad \forall M(\zeta)\eta = 0 \} \quad \forall \zeta \quad \Leftarrow \quad \exists G = -G^* : Q \prec \{M^{[\infty]*} G M^{[0]}\}^{\mathcal{H}}$ $M(\boldsymbol{\zeta}) = M^{[0]} + \boldsymbol{\zeta} M^{[\infty]}$ Positive real uncertainties, where $\{ \eta^* Q \eta < 0 \quad \forall M(\zeta) \eta = 0 \} | \forall \zeta \ge 0 | \iff \exists G = -G^* \ D \succeq 0 : Q \prec \{ M^{[\infty]*}(D+G) M^{[0]} \}^{\mathcal{H}}$ Positive complex uncertainties, where $M(\zeta) = M_0 + \zeta M_{\infty}$ $\{ \eta^* Q\eta < 0 \quad \forall M(\zeta)\eta = 0 \} \quad \forall \operatorname{Re}(\zeta) \ge 0 \quad \Leftarrow \quad \exists D \succeq 0 : Q \prec \{M^{[\infty]*} DM^{[0]}\}^{\mathcal{H}}$

• Prove $N^*(j\omega)QN(j\omega) \prec 0$ where $N(j\omega) = \begin{vmatrix} (j\omega - A)^{-1}B \\ I \end{vmatrix}$ for all $\omega \in \mathbb{R}$ **Equivalent to** $\{ \eta^* Q \eta < 0 \quad \forall [j\omega - A - B] \eta = 0 \} \quad \forall \omega$ ▲ S-variable result: $\exists G = -G^*$: $Q \prec \left\{ \begin{bmatrix} -jI \\ 0 \end{bmatrix} G \begin{bmatrix} -A & -B \end{bmatrix} \right\}^{\mathcal{H}}$ ▲ Exactly identical to KYP lemma: $\exists P = P^*(=-jG)$: $Q + \begin{bmatrix} A^*P + PA & PB \\ B^*P & 0 \end{bmatrix} \prec 0$

S-Variable - KYP lemma

• Encourages to re-examine the central control results with the S-variable / descriptor viewpoint

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• At the expense of increasing the size of π , uncertainties can be separated from design variables

$$\begin{bmatrix} E_{x} & E_{\pi} & A & B \\ 0 & KF_{\pi} & KC & KD-I \end{bmatrix} \begin{pmatrix} \dot{x} \\ \pi \\ x \\ u \end{pmatrix} = \begin{bmatrix} (1-\zeta)M^{[0]} + \zeta M^{[1]} \\ KM_{2} + M_{3} \end{bmatrix} \eta = 0 \quad \zeta \in [0 \ 1]$$

S-Variable robust stability analysis condition: $\exists P^{[v]} \succ 0, S, S^{[v]}$ such that

$$Q(P^{[v]}) \prec \left\{ SM^{[v]} + S^{[v]}(KM_2 + M_3) \right\} \quad v = 0, 1$$

S-Variable - Static-output feedback design

[Sadabadi and Peaucelle, 2016]

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$$\begin{bmatrix} E_{\mathsf{x}} & E_{\pi} & A & B\\ 0 & KF_{\pi} & KC & KD-I \end{bmatrix} \eta = \begin{bmatrix} (1-\zeta)M^{[0]} + \zeta M^{[1]}\\ KM_2 + M_3 \end{bmatrix} \eta = 0 \quad \zeta \in [0 \ 1]$$

Heuristic robust design condition: choose $K_{\pi}^{[v]}$ and $K_{x}^{[v]}$ such that the following 2 systems are stable

$$\begin{bmatrix} E_x^{[\nu]} & E_\pi^{[\nu]} & A^{[\nu]} & B^{[\nu]} \\ 0 & \mathcal{K}_\pi^{[\nu]} & \mathcal{K}_x^{[\nu]} & -I \end{bmatrix} \boldsymbol{\eta} = \begin{bmatrix} M^{[\nu]} \\ M_4^{[\nu]} \end{bmatrix} \boldsymbol{\eta} = 0 \quad \nu = 0, 1$$

any solution $K = V^{-1}W$ of the following LMIs robustly stabilizes the uncertain plant

$$Q(\mathcal{P}^{[v]}) \prec \left\{ \mathcal{SM}^{[v]} + \mathcal{M}_4^{[v]*}(\mathcal{WM}_2 + \mathcal{VM}_3)
ight\}^{\mathcal{H}} \quad v = 0, 1$$

v Good/Best choice of the full information feedback gains $K_{\pi}^{[v]}$ and $K_{x}^{[v]}$?

Choose full-information feedback $K_{\pi}^{[\nu]}$, $K_{x}^{[\nu]}$ such that the following 2 systems are stable ?

$$E_x^{[v]} = E_\pi^{[v]} + B^{[v]} \kappa_\pi^{[v]} = A^{[v]} + B^{[v]} \kappa_x^{[v]} = \hat{\eta} = 0$$
 $v = 0, 1$

S-Variable - Static full-information feedback design

and probably some more constraints (to be determined)

Possible elegant solution in [Peaucelle et al., 2024], solve

$$\hat{Q}(\mathcal{R}^{[\nu]}) \prec \left\{ \hat{M}^* \left[\begin{array}{cc} E_x^{[\nu]} \mathcal{U}_x^{[\nu]} & E_\pi^{[\nu]} \mathcal{U}_\pi^{[\nu]} + B^{[\nu]} \mathcal{W}_\pi^{[\nu]} & A^{[\nu]} \mathcal{U}_x^{[\nu]} + B^{[\nu]} \mathcal{W}_x^{[\nu]} \end{array} \right] \right\}^{\mathcal{H}} \quad \nu = 0, 1$$

where $\hat{M}\hat{\eta} = 0$ describes a stable target system and get $\mathcal{K}_{\pi}^{[\nu]} = \mathcal{W}_{\pi}^{[\nu]}(\mathcal{U}_{\pi}^{[\nu]})^{-1}$, $\mathcal{K}_{x}^{[\nu]} = \mathcal{W}_{x}^{[\nu]}(\mathcal{U}_{x}^{[\nu]})^{-1}$.



• Thank you for the good discussions and ideas!

• Looking forward for more collaborative results!



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