

Nonsmooth H_∞ Synthesis

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Outline

- ◇ context
- ◇ nonsmooth formulations
- ◇ nonsmooth descent algorithms
- ◇ applications
- ◇ concluding remarks

hard non-LMI problems

many synthesis problems do not reduce to LMI/SDP !

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- ⇒ reduced- and fixed-order synthesis (PID H_∞ , etc.)
- ⇒ structured and decentralized synthesis problems
- ⇒ general robust control with uncertain and/or nonlinear components
- ⇒ simultaneous model/controller design, multimodel control
- ⇒ unrelaxed LTI and LPV multi-objective
- ⇒ combinations of the above, etc

hard non-LMI problems

many synthesis problems do not reduce to LMI/SDP !

- ⇒ problems are nonconvex and/or nonsmooth
- ⇒ even large size LMIs/SDPs still difficult to solve

hard non-LMI problems

many synthesis problems do not reduce to LMI/SDP !

⇒ need new algorithms for hard problems

current potential techniques

- ⇒ coordinate descent schemes (D-K iterations,...) ⇒ slow and lack of convergence certificate
- ⇒ solutions based on relaxations ⇒ unknown conservatism
- other proposals
 - ⇒ TR method by Leibfritz
 - ⇒ nonsmooth gradient sampling by Burke, Lewis & Overton
 - ⇒ non-quadratic penalty method for SDP casts by Kocvara, Zibulesky,
 - ⇒ ...

fundamental limitation of SDP casts

benchmark : Boeing 767 at flutter condition

- static stabilization pb.

$$\begin{bmatrix} \dot{x} \\ y \end{bmatrix} = \begin{bmatrix} A & B \\ C & 0 \end{bmatrix} \begin{bmatrix} x \\ u \end{bmatrix}, \quad u = Ky$$

$$A = 55 \times 55 \quad B = 55 \times 2 \quad C = 2 \times 55 \quad K = 2 \times 2$$

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- ⇒ $\min_K \alpha(A + BKC) \Rightarrow 4$ variables
where $\alpha(.) := \max_i \operatorname{Re} \lambda_i(.)$ is spectral abscissa

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accept nonsmoothness and design appropriate algorithms

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$$\underset{x}{\text{minimize}} \quad \lambda_1 \left(A_0 + \sum_{i=1}^r x_i A_i + \sum_{\ell=1}^r \sum_{k=1}^s x_\ell x_k B_{\ell k} \right)$$

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- H_∞ synthesis

$$\underset{K}{\text{minimize}} \quad \sup_{\omega} \bar{\sigma} (T_{w \rightarrow z}(K, j\omega))$$

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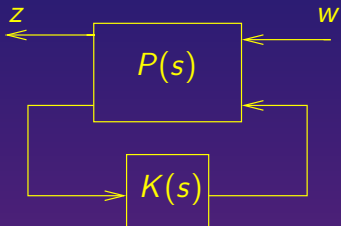
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- H_∞ synthesis

$$\underset{K}{\text{minimize}} \quad \sup_{\omega} \bar{\sigma} (T_{w \rightarrow z}(K, j\omega))$$

- many others

closed-loop H_∞ norm



$$T_{w \rightarrow z}(K(s)) := P_{11} + P_{12}K(I - P_{22}K)^{-1}P_{21}$$

- ⇒ composite $\|\cdot\|_\infty \circ T_{w \rightarrow z}$ is **Clarke regular**
- ⇒ exhaustive description of subdifferential

$$\partial(\|\cdot\|_\infty \circ T_{w \rightarrow z})(K)$$

- ⇒ variety of algorithms

Clarke subdifferential of H_∞ norm

with standard notations define closed-loop data

$$\begin{aligned}\mathcal{A}(K) &:= A + B_2 K C_2, & \mathcal{B}(K) &:= B_1 + B_2 K D_{21}, \\ \mathcal{C}(K) &:= C_1 + D_{12} K C_2, & \mathcal{D}(K) &:= D_{11} + D_{12} K D_{21},\end{aligned}$$

introduce notation

$$\begin{bmatrix} T_{w \rightarrow z}(K, s) & G_{12}(K, s) \\ G_{21}(K, s) & \star \end{bmatrix} := \begin{bmatrix} \mathcal{C}(K) \\ C_2 \end{bmatrix} (sI - \mathcal{A}(K))^{-1} \begin{bmatrix} B(K) & B_2 \end{bmatrix} + \begin{bmatrix} D(K) & D_{12} \\ D_{21} & \star \end{bmatrix}.$$

Clarke subdifferential of H_∞ norm

subdif. is convex set of subgradients Φ_Y 's

$$\frac{\sum_{\nu=1}^p \operatorname{Re} \left\{ G_{21}(K, j\omega_\nu) T_{w \rightarrow z}(K, j\omega_\nu)^H Q_\nu Y_\nu Q_\nu^H G_{12}(K, j\omega_\nu) \right\}^T}{\|T_{w \rightarrow z}(K)\|_\infty^{-1}}$$

- ⇒ ω_ν are active (peak) frequencies at K
- ⇒ $Y := (Y_1, \dots, Y_p)$ ranges over (convex)

$$\{Y = (Y_1, \dots, Y_p) : Y_i = Y_i^H, \succeq 0, \sum_{\nu=1}^p \operatorname{Tr}(Y_\nu) = 1\}$$

convergent nonsmooth descent method

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- ⇒ define descent function (1st-order model)

$$\theta(K) := \inf_{H \in \mathbb{R}^{m_2 \times p_2}} \sup_{\omega \in \Omega_e(K)} \sup_{\text{Tr } Y_\omega = 1, Y_\omega \succeq 0} \left\{ \dots \right. \\ \left. - \|T_{w \rightarrow z}(K)\|_\infty + \bar{\sigma}(T_{w \rightarrow z}(K, j\omega)) + \langle \Phi_{Y_\omega}, H \rangle + \frac{1}{2} \delta \|H\|_F^2 \right\}$$

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- ⇒ H is (controller) search direction computed explicitly

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⇒ Fenchel duality $\implies$ descent direction at K

$$H(K) := -\frac{1}{\delta} \sum_{\omega \in \Omega_e(K)} \tau_\omega \Phi_{Y_\omega}.$$

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$$\begin{aligned} \theta(K) &:= \sup_{\sum_{\omega \in \Omega_e(K)} \tau_\omega = 1, \tau_\omega \geq 0} \sup_{Y_\omega = 1, Y_\omega \succeq 0} \left\{ -\|T_{w \rightarrow z}(K)\|_\infty \dots \right. \\ &\quad \left. + \sum_{\omega \in \Omega_e(K)} \tau_\omega \bar{\sigma}(T_{w \rightarrow z}(K, j\omega)) - \frac{1}{2\delta} \left\| \sum_{\omega \in \Omega_e(K)} \tau_\omega \Phi_{Y_\omega} \right\|_F^2 \right\} \end{aligned}$$

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⇒ $\theta(K) \leq 0, \forall K$ and $\theta(K) = 0 \Leftrightarrow 0 \in \partial(\|\cdot\|_\infty \circ T_{w \rightarrow z})(K)$

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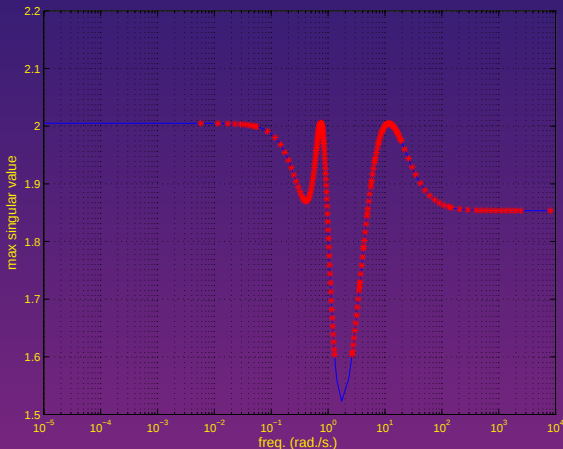
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⇒ **algorithm**: line search using $H(K)$ with $\theta(K)$ as stopping test

example of enriched set $\Omega_e(K)$

- $\Omega_e(K)$ captures extra info. on $\bar{\sigma}(T_{w \rightarrow z}(K, j\omega)) \implies$ better steps



fixed-order output-feedback H_∞ synthesis

examples from Leibfritz's collection

problem	(n, m, p)	order	iter	cpu (sec.)	nonsmooth H_∞	H_∞ AL	FW	H_∞ full
AC8	(9, 1, 5)	0	20	45	2.005	2.02	2.612	1.62
HE1	(4, 2, 1)	0	4	7	0.154	0.157	0.215	0.073
REA2	(4, 2, 2)	0	31	51	1.192	1.155	1.263	1.141
AC10	(55, 2, 2)	0	15	294	13.11	*	*	3.23
AC10	(55, 2, 2)	1	46	408	10.21	*	*	3.23
BDT2	(82, 4, 4)	0	44	1501	0.8364	*	*	0.2340
HF1	(130, 1, 2)	0	11	1112	0.447	*	*	0.447
CM4	(240, 1, 2)	0	2	3052	0.816	*	*	*
CM5	(480, 1, 2)	0	2	4785	0.816	*	*	*

H_∞ synthesis with nonsmooth algorithm
low-level SUNW, Sun-Blade-1500

extensions

⇒ pure stabilization problems $A + B_2 K C_2$ is Hurwitz iff

$$\|C_2(sI - (A + B_2 K C_2))^{-1} B_2\|_\infty < \infty$$

⇒ multidisk problems

$$\underset{K}{\text{minimize}} f(K) := \max_{i=1,\dots,N} \|T_{w^i \rightarrow z^i}(K)\|_\infty$$

⇒ structured feedback design

$$K = K_0 + L \text{diag}(\kappa) R$$

⇒ etc

concluding remarks

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- ⇒ promising for IQC and μ synthesis

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- ⇒ easily extended to any controller structure
- ⇒ convergence certificate
- ⇒ can handle large state systems
- ⇒ promising for IQC and μ synthesis
- ⇒ second-order version in progress