# Application of structured linearization for efficient $\mathcal{H}_{\infty}$-norm computation 

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## Outline

(1) $\mathcal{H}_{\infty}$-norm definition
(2) Current method

- New method (using 'loss of dissipativity' property)
- Bezoutian matrix
© Time improvement


## H-infinity norm

For a transfer matrix $G(s) \in \mathbb{R}(s)^{p \times m}$ (with $p$ rows and $m$ columns, entries from $\mathbb{R}(s))$ :

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\|G\|_{\mathcal{H}_{\infty}}:=\sup _{\lambda \in \mathbb{C}, \operatorname{Real}(\lambda) \geqslant 0} \sigma_{\max }(G(\lambda))
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where $\sigma_{\max }(P)$ is the maximum singular value of a constant matrix $P$.

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This talk: about computation of $\|G\|_{\mathcal{H}_{\infty}}$

## Conventional method to compute $\|G\|_{\mathcal{H}_{\infty}}$

Obtain a state space realization of $G(s):(A, B, C, D)$
Assuming $\gamma>\sigma_{\max }(D)$

Define: $\quad H(\gamma):=\left[\begin{array}{cc}A-B R^{-1} D^{T} C & -\gamma B R^{-1} B^{T} \\ \gamma C^{T} S^{-1} C & -A^{T}+C^{T} D R^{-1} B^{T}\end{array}\right]$
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\gamma \leqslant\|G\|_{\mathcal{H}_{\infty}} \Leftrightarrow H(\gamma) \text { has } i \mathbb{R} \text { eigenvalues }
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to iterate over $\gamma$ to find $\gamma=\|G\|_{\mathcal{H}_{\infty}}$ to required accuracy. For each $\gamma$ value: solve an eigenvalue problem

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Thus
$\|G\|_{\mathcal{H}_{\infty}}=\min \{\gamma \mid \gamma$ satisfies $P(\gamma, \omega) \geqslant 0$ as above $\}$
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Look for those $\gamma$ 's that make $p_{\gamma}(\omega)$ and $\frac{\partial p}{\partial \omega}=: q_{\gamma}(\omega)$ noncoprime Discriminant / Resultant results applicable now Bezoutian matrix better than Sylvester matrix

## Bezoutian of two polynomials

Consider polynomials $p(\omega)$ and $q(\omega)$.
Bezoutian polynomial $b(\zeta, \eta)$ and matrix $B$ are defined as

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b(\zeta, \eta):=\frac{p(\zeta) q(\eta)-p(\eta) q(\zeta)}{\zeta-\eta}=\left[\begin{array}{c}
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$B$ is a symmetric matrix
But not the Sylvester resultant matrix $S$

## Construction of $b(\zeta, \eta)$ and $B$

Let $p(\xi)=p_{0}+p_{1} \xi+p_{2} \xi^{2} \cdots+p_{n} \xi^{n}$ and $q(\xi)=q_{0}+q_{1} \xi+q_{2} \xi^{2} \cdots+q_{m} \xi^{m}$ and $S$, the Sylvester resultant matrix is $(n+m) \times(n+m)$

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But $S$ is quite 'abnormal' ( $A$ is called normal if $A A^{T}=A^{T} A$ ) At least the zero eigenvalue of $S$ (when $S$ is singular) is ill-conditioned.
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But Bezoutian matrix: symmetric

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Degree of $b(\zeta, \eta)$ is $n-1$ in each of $\zeta$ and $\eta$ and matrix $B$ is $n \times n$. $b(\zeta, \eta)$ explicit calculation quite easy to program (in Scilab): Let $b(\zeta, \eta)=b_{0}(\zeta)+\eta b_{1}(\zeta)+\eta^{2} b_{2}(\zeta) \cdots \eta^{n-1} b_{n-1}(\zeta)$
Equate terms with equal degree in $\eta$ in

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(\zeta-\eta) b(\zeta, \eta)=p(\zeta) q(\eta)-q(\zeta) p(\eta) \text { to get }
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## Example

Let $p(s)=2+3 s+s^{2}$ and $q_{a}(s)=a+s$
Sylvester resultant matrix $S=\left[\begin{array}{lll}2 & 3 & 1 \\ a & 1 & 0 \\ 0 & a & 1\end{array}\right]$ $\operatorname{det} S=a^{2}-3 a+2$ :

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$B_{a}=\left[\begin{array}{cc}1 & a \\ a & (3 a-2)\end{array}\right]$ with determinant $-\left(a^{2}-3 a+2\right)$

## $\mathcal{H}_{\infty}$ norm computation: Bezoutian polynomial

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Hamiltonian matrix $H(\gamma)$ has imaginary axis eigenvalues. Recall that

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\gamma \leqslant\|G\|_{\mathcal{H}_{\infty}} \Leftrightarrow H(\gamma) \text { has } i \mathbb{R} \text { eigenvalues }
$$

## Structured linearization

Define symmetric matrices $E$ and $A$ as

$$
E:=\left[\begin{array}{cccc}
B_{m} & -B_{m-2} & \cdots & -B_{1} \\
\hline & -B_{0} \\
\vdots & . & . & 0 \\
-B_{1} & \because & & 0 \\
-B_{0} & 0 & &
\end{array}\right] \quad A:=\left[\begin{array}{cccc}
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- But need to check that common root is on $i \mathbb{R}$, maximum such $\gamma$, etc.
Program implemented in Scilab (and code on github)


## Comparison of two methods (large order)

Plot of time taken by the two methods (strictly proper case)


Figure: Plot from Belur \& Praagman, IEEE-TAC, 2011

## Comparison of two methods (small order)



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## Conclusion

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- But requires an eigenvalue iteration within $\gamma$ iteration
- Sylvester matrix resultant (and block companion linearization): marginally better
- Proposed method is much faster due to symmetric linearization:
20 to 40 times faster. Improvement further better for higher orders


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