On A Quadratic Eigenproblem Arising In The Analysis of Delay Equations

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Joint work with E. Jarlebring, N. & D.S. Mackey

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Polynomial Matrix Eigenproblem Spectral Symmetry Cayley Transformations Structured Linearization Conclusions

TDS Critical System Quadratic eigenproblem

Outline

- Time Delay System
- Polynomial Eigenvalue Problem
- Spectral Symmetry
- Structured Linearization
- Conclusions

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Time Delay Systems (TDS)

$$\dot{x}(t) = A_0 x(t) + \sum_{k=1}^{m} A_k x(t - h_k), \ t > 0$$

 $x(t) = \varphi(t), \ t \in [-h_m, 0]$ (Σ)

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with $0 < h_1 < \ldots < h_m$ and $A_k \in \mathbb{R}^{n \times n}$.

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with $0 < h_1 < \ldots < h_m$ and $A_k \in \mathbb{R}^{n \times n}$.

Definition

• Eigenvalue s and eigenvector $v \neq 0$:

$$\mathbb{M}(s)\mathbf{v} := \left(-sI_n + A_0 + \sum_{k=1}^m A_k e^{-h_k s}\right)\mathbf{v} = 0$$

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• spectrum $\sigma(\Sigma)$: set of all eigenvalues

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- spectrum $\sigma(\Sigma)$: set of all eigenvalues
- stable: $\sigma(\Sigma) \subset \mathbb{C}^-$

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Critical System

Problem

For what h_1, \ldots, h_m is there an ω s.t

 $\mathbb{M}(\imath\omega)\mathbf{v}=\mathbf{0}.$

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Definition

 Σ is called critical iff $\sigma(\Sigma) \cap i\mathbb{R} \neq \emptyset$.

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Example (Jarlebring 2005)

Two delay problem: $\dot{x}(t) = -x(t - h_1) - 2x(t - h_2)$

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Example (Jarlebring 2005)

Two delay problem: $\dot{x}(t) = -x(t - h_1) - 2x(t - h_2)$ Critical curves:



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For what h_1, \ldots, h_m is there an ω s.t

 $\mathbb{M}(\imath\omega)\mathbf{v}=\mathbf{0}.$

Hale & Huang 1993: Scalar two delays: Geometric classification Chen & Gu & Nett 1995: Commensurate delays Louisell 2001: Single delay, neutral, moderate size Sipahi & Olgac 2003 : Small systems, few delays: Form determinant + Routh table + Rekasius Substitution.

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Given free parameters φ_k , $k = 1, \ldots, m - 1$.

Theorem (Jarlebring 2005)

The point in delay space (h_1, \ldots, h_m) is critical iff

$$h_{k} = \frac{\varphi_{k} + 2p\pi}{\omega}, \ k = 1, \dots, m-1$$
$$h_{m} = \frac{\operatorname{Arg} s + 2q\pi}{\omega}$$

$$\left[s^2 I \otimes A_m + s \left(\sum_{k=0}^{m-1} I \otimes A_k e^{-i\varphi_k} + e^{i\varphi_k} A_k \otimes I\right) + A_m \otimes I\right] u = 0,$$

where $s = e^{\imath \omega}$, $u = \text{vec } vv^* = v \otimes \overline{v}$ and

$$\omega = -\imath v^* \left(A_0 + \sum_{k=1}^{m-1} A_k e^{-\imath \varphi_k} + A_m s \right) v.$$

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Quadratic eigenproblem

$$\left[s^2(I\otimes A_m)+s\left(\sum_{k=0}^{m-1}I\otimes A_ke^{-i\varphi_k}+e^{i\varphi_k}A_k\otimes I\right)+(A_m\otimes I)\right]u=0,$$

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Quadratic eigenproblem



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Quadratic Eigenvalue Problem

$$M \in \mathbb{R}^{n^2 \times n^2}$$
$$G \in \mathbb{C}^{n^2 \times n^2}$$
$$K \in \mathbb{R}^{n^2 \times n^2}$$

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Theorem (Horn, Johnson)

There exists an involutary permutation matrix $P \in \mathbb{R}^{n^2 \times n^2}$ such that $B \otimes C = P(C \otimes B)P$ for all $B, C \in \mathbb{R}^{n \times n}$.

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In particular,

$$P = \sum_{i,j=1}^{n} E_{ij} \otimes E_{ij}^{T} = [E_{ij}^{T}]_{i,j=1}^{n},$$

where $E_{ij} \in \mathbb{R}^{n \times n}$ has entry 1 in position i, j and all other entries are zero.

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where $E_{ij} \in \mathbb{R}^{n \times n}$ has entry 1 in position i, j and all other entries are zero. $\begin{bmatrix} 1 & 0 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} 0 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} 0 & 0 & 0 & 0 \end{bmatrix}$

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Hence, we have

$$M = A_m \otimes I = P(I \otimes A_m)P = PKP,$$

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There exists an involutary permutation matrix $P \in \mathbb{R}^{n^2 \times n^2}$ such that $B \otimes C = P(C \otimes B)P$ for all $B, C \in \mathbb{R}^{n \times n}$.

Hence, we have

$$M = A_m \otimes I = P(I \otimes A_m)P = PKP,$$

and

$$A_k \otimes I = P(I \otimes A_k)P$$

such that

$$G = \sum_{k=0}^{m-1} e^{-i\varphi_k} (I \otimes A_k) + e^{i\varphi_k} (A_k \otimes I)$$

= $P\left(\sum_{k=0}^{m-1} (A_k \otimes I) e^{-i\varphi_k} + e^{i\varphi_k} (I \otimes A_k)\right) P = P\overline{G}P.$

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As M and K are real, this implies

$Q(z) = z^2 M + zG + K = z^2 P K P + z P \overline{G} P + P M P = P(z^2 \overline{K} + z \overline{G} + \overline{M}) P,$

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TDS Critical System Quadratic eigenproblem

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that is, Q(z) is a matrix polynomial which satisfies

 $Q = P \cdot \operatorname{rev}(\overline{Q}) \cdot P,$

with

$$\overline{Q}(z)=z^2\overline{M}+z\overline{G}+\overline{K},$$

and

$$\operatorname{rev}(Q(z)) := z^2 Q(\frac{1}{z}) = M + zG + z^2 K.$$

Problem Statement

Problem Statement

We will consider

$$Q(\lambda)v = 0 ext{ with } Q(\lambda) = \sum_{i=0}^k \lambda^i B_i, ext{ } B_k
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$$Q(\lambda)v = 0$$
 with $Q(\lambda) = \sum_{i=0}^{k} \lambda^{i} B_{i}, \qquad B_{k} \neq 0, \quad B_{i} \in \mathbb{C}^{n \times n},$

which satisfies

$$Q(\lambda) = P \cdot \operatorname{rev}(\overline{Q}(\lambda)) \cdot P$$

for an involutary permutation matrix P.

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As
$$Q(\lambda) = \sum_{i=0}^{k} \lambda^{i} B_{i}$$
, this implies $B_{i} = P\overline{B}_{k-i}P$, $i = 0, \dots, k$.

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$$Q(\lambda) = \sum_{i=0}^{k} \lambda^{i} B_{i}$$
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Questions to be answered:

- eigenvalue pairing
- structured linearizations

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Problem Statement

Structure reminds of:

- (anti-)palindromic: $\pm \operatorname{rev}(\mathcal{Q}(\lambda)) = \mathcal{Q}(\lambda)$
- *-(anti-)palindromic: $\pm \operatorname{rev}(Q^*(\lambda)) = Q(\lambda)$
- even, odd: $\pm Q(-\lambda) = Q(\lambda)$
- *-even, *-odd: $\pm Q^*(-\lambda) = Q(\lambda)$

where \star is used for transpose T in the real case and either T or conjugate transpose * in the complex case.

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- *-even, *-odd: $\pm Q^{\star}(-\lambda) = Q(\lambda)$

where \star is used for transpose T in the real case and either T or conjugate transpose * in the complex case.

Recall

 $Q(\lambda) = \pm P \cdot \operatorname{rev}(\overline{Q}(\lambda)) \cdot P.$

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Recall

 $Q(\lambda) = \pm P \cdot \operatorname{rev}(\overline{Q}(\lambda)) \cdot P.$

Define even/odd equivalent

$$Q(\lambda) = \pm P \cdot \overline{Q}(-\lambda) \cdot P.$$

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Recall PCP-(anti-)palindromic (short PCP/anti-PCP)

 $Q(\lambda) = \pm P \cdot \operatorname{rev}(\overline{Q}(\lambda)) \cdot P.$

Define even/odd equivalent PCP-even/odd

$$Q(\lambda) = \pm P \cdot \overline{Q}(-\lambda) \cdot P.$$

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Spectral Symmetry

Let $Q(\lambda)v = 0$, and Q is PCP, then we have

$$0 = Q(\lambda)v = P \cdot \operatorname{rev}(\overline{Q}(\lambda)) \cdot Pv$$

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Spectral Symmetry

Let $Q(\lambda)v = 0$, and Q is PCP, then we have

$$0 = Q(\lambda)v = P \cdot \operatorname{rev}(\overline{Q}(\lambda)) \cdot Pv$$

which implies

$$\operatorname{rev}(\overline{Q}(\lambda)) \cdot (Pv) = 0$$

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Spectral Symmetry

Let $Q(\lambda)v = 0$, and Q is PCP, then we have

$$0 = Q(\lambda)v = P \cdot \operatorname{rev}(\overline{Q}(\lambda)) \cdot Pv$$

which implies

$$\operatorname{rev}(\overline{Q}(\lambda)) \cdot (Pv) = 0$$

and

$$Q(1/\overline{\lambda})\cdot(P\overline{v})=0.$$

Hence, if λ is an eigenvalue with eigenvector v, then $1/\overline{\lambda}$ is an eigenvalue with eigenvector $P\overline{v}$.

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Theorem

Let $Q(\lambda) = \sum_{i=0}^{k} \lambda^{i} B_{i}$, $B_{k} \neq 0$ be a regular matrix polynomial, that is, det $Q(\lambda)$ is not identically zero for all $\lambda \in \mathbb{C}$.

- If $Q(\lambda) = \pm P \cdot \operatorname{rev}(\overline{Q}(\lambda)) \cdot P$, then the spectrum of $Q(\lambda)$ has the eigenvalue pairing $(\lambda, 1/\overline{\lambda})$.
- **3** If $Q(\lambda) = \pm P \cdot \overline{Q}(-\lambda) \cdot P$, then the spectrum of $Q(\lambda)$ has the eigenvalue pairing $(\lambda, -\overline{\lambda})$

Moreover, the algebraic, geometric, and partial multiplicities of the two eigenvalues in each such pair are equal. (Here, we allow $\lambda = 0$ and interpret $1/\lambda$ as the eigenvalue ∞ .)

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- **2** If $Q(\lambda) = \pm P \cdot \overline{Q}(-\lambda) \cdot P$, then the spectrum of $Q(\lambda)$ has the eigenvalue pairing $(\lambda, -\overline{\lambda})$

Moreover, the algebraic, geometric, and partial multiplicities of the two eigenvalues in each such pair are equal. (Here, we allow $\lambda = 0$ and interpret $1/\lambda$ as the eigenvalue ∞ .)

Idea of the proof of statement 1: $Q(\lambda)$ and its first companion form $C_1(\lambda) = \lambda X + Y$ have the same eigenvalues (including multiplicities). C_1 of a (anti-)PCP Q is strictly equivalent to $X^* + \lambda Y^*$.

Structure of $Q(\lambda)$	eigenvalue pairing
(anti)-palindromic, T-(anti)-palindromic	$(\lambda, 1/\lambda)$
*-palindromic, *-anti-palindromic	$(\lambda,1/\overline{\lambda})$
(anti)-PCP-palindromic	$(\lambda, 1/\overline{\lambda})$
even, odd, T-even, T-odd	$(\lambda, -\lambda)$
*-even, *-odd	$(\lambda,-\overline{\lambda})$
PCP-even, PCP-odd	$(\lambda, -\overline{\lambda})$

Spectral symmetries

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Cayley Transformations

The Cayley transformation for a matrix polynomial $Q(\lambda)$ of degree k with pole at +1 or -1, resp., is

$$egin{array}{rll} \mathcal{C}_{+1}(Q)(\mu) &:= & (1-\mu)^k Q(rac{1+\mu}{1-\mu}), \ \mathcal{C}_{-1}(Q)(\mu) &:= & (\mu+1)^k Q(rac{\mu-1}{\mu+1}). \end{array}$$

Cayley Transformations Structured Linearization

	$\mathcal{C}_{-1}(Q)(\mu) = (\mu+1)^k Q(rac{\mu-1}{\mu+1})$	
$Q(\lambda)$	k even	k odd
palindromic	even	odd
\star -palindromic	∗-even	*-odd
anti-palindromic	odd	even
\star -anti-palindromic	*-odd	∗-even
PCP	PCP-even	PCP-odd
anti-PCP	PCP-odd	PCP-even
even	palindromic	
∗-even	*-palindromic	
odd	anti-palindromic	
*-odd	\star -anti-palindromic	
PCP-even	РСР	
PCP-odd	anti-PCP	

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	${\mathcal C}_{+1}(Q)(\mu) = (1-\mu)^k Q(rac{1+\mu}{1-\mu})$		
$Q(\lambda)$	k even	k odd	
palindromic	even		
\star -palindromic	∗ -even		
anti-palindromic	odd		
\star -anti-palindromic	*-odd		
PCP	PCP-even		
anti-PCP	PCP-odd		
even	palindromic	anti-palindromic	
∗-even	\star -palindromic	∗-anti-palindromic	
odd	anti-palindromic	palindromic	
*-odd	\star -anti-palindromic	\star -palindromic	
PCP-even	PCP	anti-PCP	
PCP-odd	anti-PCP	PCP	

Cavley transformations

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On A Quadratic Eigenproblem Arising In The Analysis of Delay E

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Linearization Where to find? Structured PCP-Pencil Structured Linearization Structured PCP-even/odd-Linearization

Linearization

The classical approach to solve $Q(\lambda)v = 0$ for

$$Q(\lambda) = \sum_{i=0}^k \lambda^i B_i, \qquad B_k
eq 0$$

is linearization, in which the given polynomial is transformed into a $kn \times kn$ matrix pencil $L(\lambda) = \lambda X + Y$ that satisfies

$$E(\lambda)L(\lambda)F(\lambda) = \begin{bmatrix} Q(\lambda) & 0\\ 0 & I_{(k-1)n} \end{bmatrix},$$

where $E(\lambda)$ and $F(\lambda)$ are unimodular matrix polynomials. (A matrix polynomial is unimodular if its determinant is a nonzero constant, independent of λ).

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Let
$$X_1 = X_2 = \operatorname{diag}(B_k, I_n, \ldots, I_n)$$
,

$$Y_{1} = \begin{bmatrix} B_{k-1} & B_{k-2} & \cdots & B_{0} \\ -I_{n} & 0 & \cdots & 0 \\ & \ddots & \ddots & \vdots \\ 0 & & -I_{n} & 0 \end{bmatrix}, \quad Y_{2} = \begin{bmatrix} B_{k-1} & -I_{n} & 0 \\ B_{k-2} & 0 & & \\ \vdots & \vdots & \ddots & -I_{n} \\ B_{0} & 0 & \cdots & 0 \end{bmatrix}$$

Then $C_1(\lambda) = \lambda X_1 + Y_1$ and $C_2(\lambda) = \lambda X_2 + Y_2$ are the first and second companion forms for $Q(\lambda)$. These linearizations do not reflect the structure present in the matrix polynomial Q.

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Structured Linearization

Source of linearizations: [Mackey, Mackey, Mehl, Mehrmann 2006]

$$\begin{split} \mathbb{L}_{1}(Q) &= \left\{ L(\lambda) = \lambda X + Y : L(\lambda) \cdot (\Lambda \otimes I_{n}) = v \otimes Q(\lambda), v \in \mathbb{C}^{k} \right\}, \\ \mathbb{L}_{2}(Q) &= \left\{ L(\lambda) = \lambda X + Y : (\Lambda^{T} \otimes I_{n}) \cdot L(\lambda) = w^{T} \otimes Q(\lambda), w \in \mathbb{C}^{k} \right\} \\ \text{where} \qquad \Lambda = [\lambda^{k-1} \ \lambda^{k-2} \ \cdots \ \lambda \ 1]^{T}. \end{split}$$

v is called right ansatz vector, w left ansatz vector.

$$\dim \mathbb{L}_1(Q) = \dim \mathbb{L}_2(Q) = k(k-1)n^2 + k$$

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Structured PCP-Pencil

We have Q which satisfies

$$P \cdot \operatorname{rev}(\overline{Q}(\lambda)) \cdot P = Q(\lambda)$$

for some $n \times n$ real involution P.

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$$P \cdot \operatorname{rev}(\overline{Q}(\lambda)) \cdot P = Q(\lambda)$$

for some $n \times n$ real involution P.

We want a pencil $L(\lambda) \in \mathbb{L}_1(Q)$ such that

$$\widehat{P} \cdot \operatorname{rev}(\overline{L}(\lambda)) \cdot \widehat{P} = L(\lambda)$$

for some $kn \times kn$ real involution \widehat{P} .

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We want a pencil $L(\lambda) \in \mathbb{L}_1(Q)$ such that

$$\widehat{P} \cdot \operatorname{rev}(\overline{L}(\lambda)) \cdot \widehat{P} = L(\lambda)$$

for some $kn \times kn$ real involution \widehat{P} . It is not immediately obvious what to use for \widehat{P} .

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Back to

$$(\lambda^2 M + \lambda G + K)v = 0,$$

with

 $M = P\overline{K}P$ and $G = P\overline{G}P$.

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$$\widehat{P} = \left[\begin{array}{c} P \\ P \end{array} \right]$$

does not work as there are no pencils in $\mathbb{L}_1(Q)$ satisfying

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unless the matrix G is very specifically tied to the leading coefficient M,

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$$\widehat{P} \cdot \operatorname{rev}(\overline{L}(\lambda)) \cdot \widehat{P} = L(\lambda),$$

unless the matrix G is very specifically tied to the leading coefficient M, e.g. for $v = \begin{bmatrix} 1 & 1 \end{bmatrix}^T$

$$G=P\overline{M}P+M=K+M.$$

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Choosing

$$\widehat{P} = \left[\begin{array}{c} P \\ P \end{array} \right]$$

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Choosing

$$\widehat{P} = \left[\begin{array}{c} P \\ P \end{array} \right]$$

restricts the ansatz vector $v = [v_1 \ v_2]^T \in \mathbb{C}^2$ to

$$R_2 v = \overline{v}$$
 with $R_2 = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$,

that is $v_1 = \overline{v}_2$

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that is $v_1 = \overline{v}_2$ and

$$\lambda X + Y = \lambda \begin{bmatrix} v_1 M & -W_1 \\ \overline{v}_1 M & \overline{v}_1 G + P \overline{W}_1 P \end{bmatrix} + \begin{bmatrix} W_1 + v_1 G & v_1 P \overline{M} P \\ -P \overline{W}_1 P & \overline{v}_1 P \overline{M} P \end{bmatrix},$$

where W_1 is arbitrary, is a structured pencil in $\mathbb{L}_1(Q)$.

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Structured Linearization

For regular Q and $L(\lambda) \in \mathbb{L}_1(Q)$ with $v \neq 0, v \in \mathbb{C}^2$

- select any nonsingular matrix T such that $Tv = \alpha e_1$
- apply $T \otimes I_n$ to $L(\lambda)$ to produce $\widetilde{L}(\lambda) = (T \otimes I_n) \cdot L(\lambda)$

$$\widetilde{L}(\lambda) = \lambda \widetilde{X} + \widetilde{Y} = \lambda \begin{bmatrix} \widetilde{X}_{11} & \widetilde{X}_{12} \\ 0 & -Z \end{bmatrix} + \begin{bmatrix} \widetilde{Y}_{11} & \widetilde{Y}_{12} \\ Z & 0 \end{bmatrix},$$

where \widetilde{X}_{11} and \widetilde{Y}_{12} are $n \times n$.

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where \widetilde{X}_{11} and \widetilde{Y}_{12} are $n \times n$.

If det $Z \neq 0$, $L(\lambda)$ is a linearization of Q. [Mackey, Mackey, Mehl, Mehrmann 2006]

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As $v = \begin{bmatrix} v_1 & \overline{v}_1 \end{bmatrix}^T$ choose T as

$$T = \left[\begin{array}{cc} \overline{v}_1 & v_1 \\ -\overline{v}_1 & v_1 \end{array} \right].$$

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This yields

$$-Z = |v_1|^2 G + \overline{v}_1 W_1 + v_1 P \overline{W}_1 P.$$

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As $v = \begin{bmatrix} v_1 & \overline{v}_1 \end{bmatrix}^T$ choose T as

$$\mathcal{T} = \left[egin{array}{cc} \overline{\mathbf{v}}_1 & \mathbf{v}_1 \ -\overline{\mathbf{v}}_1 & \mathbf{v}_1 \end{array}
ight].$$

This yields

$$-Z = |v_1|^2 G + \overline{v}_1 W_1 + v_1 P \overline{W}_1 P.$$

As $Q(\lambda) = \lambda^2 M + \lambda G + K$ is regular, we have for $W_1 = v_1 M$

$$-Z = |v_1|^2(G + M + P\overline{M}P) = |v_1|^2(G + M + K) = |v_1|^2Q(1),$$

and det $Z \neq 0$ iff 1 is not an eigenvalue of Q.

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Other possible choice of W_1 :

- $W_1 = v_1 M$ yields det $Z \neq 0$ if 1 is not an eigenvalue of Q.
- 3 $W_1 = -v_1 M$ yields det $Z \neq 0$ if -1 is not an eigenvalue of Q.
- **3** $W_1 = \overline{v}_1 M$ yields det $Z \neq 0$ if $\frac{\overline{v}_1}{v_1}$ is not an eigenvalue of Q.
- $W_1 = -\overline{v}_1 M$ yields det $Z \neq 0$ if $-\frac{\overline{v}_1}{v_1}$ is not an eigenvalue of Q.
- $W_1 = v_1 G$ yields det $Z \neq 0$ if det $G \neq 0$.

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For $W_1 = -\overline{v}_1 M$ we have

$$\lambda \left[\begin{array}{cc} v_1 M & \overline{v}_1 M \\ \overline{v}_1 M & \overline{v}_1 G - v_1 P \overline{M} P \end{array} \right] + \left[\begin{array}{cc} v_1 G - \overline{v}_1 M & v_1 P \overline{M} P \\ v_1 P \overline{M} P & \overline{v}_1 P \overline{M} P \end{array} \right] \in \mathbb{L}_1(Q) \cap \mathbb{L}_2(Q)$$

if $\frac{\overline{v}_1}{v_1}$ is not an eigenvalue of Q.

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Linearization Where to find? Structured PCP-Pencil **Structured Linearization** Structured PCP-even/odd-Linearization

For $W_1 = -\overline{v}_1 M$ we have

$$\lambda \left[\begin{array}{cc} \mathsf{v}_1 M & \overline{\mathsf{v}}_1 M \\ \overline{\mathsf{v}}_1 M & \overline{\mathsf{v}}_1 G - \mathsf{v}_1 P \overline{M} P \end{array} \right] + \left[\begin{array}{cc} \mathsf{v}_1 G - \overline{\mathsf{v}}_1 M & \mathsf{v}_1 P \overline{M} P \\ \mathsf{v}_1 P \overline{M} P & \overline{\mathsf{v}}_1 P \overline{M} P \end{array} \right] \in \mathbb{L}_1(Q) \cap \mathbb{L}_2(Q)$$

if $\frac{\overline{v}_1}{v_1}$ is not an eigenvalue of Q.

Similar construction for general (anti-)PCP-polynomial possible.

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Linearization Where to find? Structured PCP-Pencil Structured Linearization Structured PCP-even/odd-Linearization

Structured PCP-even/odd-Linearization

We have Q which satisfies

$$\pm P \cdot \overline{Q}(-\lambda) \cdot P = Q(\lambda)$$

for some $n \times n$ real involution P.

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Structured PCP-even/odd-Linearization

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We want a pencil $L(\lambda) \in \mathbb{L}_1(Q)$ such that

$$\pm \widehat{P} \cdot \overline{L}(-\lambda) \cdot \widehat{P} = L(\lambda)$$

for some $2n \times 2n$ real involution \widehat{P} .

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Structured PCP-even/odd-Linearization

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We want a pencil $L(\lambda) \in \mathbb{L}_1(Q)$ such that

$$\pm \widehat{P} \cdot \overline{L}(-\lambda) \cdot \widehat{P} = L(\lambda)$$

for some $2n \times 2n$ real involution \widehat{P} . It is not immediately obvious what to use for \widehat{P} .

Introduction Polynomial Matrix Eigenproblem Spectral Symmetry Cayley Transformations Structured Linearization Conclusions Structured PCP-Pencil Structured Linearization Structured PCP-even/odd-Linearization

Neither

$$\widehat{P} = \begin{bmatrix} P & \\ P \end{bmatrix}$$
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nor

work unless the coefficient matrices of Q are very specifically tied to each other.

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does the job.

Construction for general PCP-even/odd polynomial possible.

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- New Structured Polynomial Eigenvalue Problem
- Spectral Symmetry
- Cayley Transformation
- Structured Linearization for (anti-)PCP and PCP-even/odd polynomials

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- New Structured Polynomial Eigenvalue Problem
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Open Problems:

- Choice of the Ansatz Vector v
- Structure-Preserving Transformation
- Structure-Preserving Eigenvalue Algorithm

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GAMM Activity Group Meeting:

Today, 1:20 - 2:20 pm

Members as well as non-members are invited!

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