Distance bounds for prescribed multiple eigenvalues of matrix polynomials

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Introduction and preliminaries

Wilkinson's problem (1972) concerns computing the spectral norm distance from a matrix $A \in \mathbb{C}^{n \times n}$ with n distinct eigenvalues to the set of $n \times n$ matrices having multiple eigenvalues, and has a strong connection to ill-conditioning of eigenvalue problems.

Malyshev (1999) provided a solution to Wilkinson's problem by obtaining

 $\inf \{ \|E\|_2 : \mu \text{ is a multiple eigenvalue of } A + E \}$

$$= \sup_{\gamma>0} s_{2n-1} \left(\begin{bmatrix} I\mu - A & 0\\ \gamma I & I\mu - A \end{bmatrix} \right),$$

where $s_1(\cdot) \ge s_2(\cdot) \ge s_3(\cdot) \ge \cdots$ denote the singular values of a matrix.

Mengi (2011) derived a characterization for the smallest perturbation to a matrix with an eigenvalue of specified algebraic multiplicity by proving

 $\inf \{ \|E\|_2 : \mu \text{ is an eigenvalue of } A + E \text{ of algebraic multiplicity } r \}$

$$= \sup_{\gamma_{i,j} \in \mathbb{C} \setminus \{0\}} s_{rn-r+1} \left(\begin{bmatrix} I\mu - A & 0 & 0 & \cdots & 0 \\ \gamma_{2,1}I & I\mu - A & 0 & \cdots & 0 \\ \gamma_{3,1}I & \gamma_{3,2}I & I\mu - A & \cdots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ \gamma_{r,1}I & \gamma_{r,2}I & \gamma_{r,3}I & \cdots & I\mu - A \end{bmatrix} \right).$$

In this work, we study the case of polynomial eigenvalue problems, and estimate a weighted distance from a given $n \times n$ matrix polynomial to the $n \times n$ matrix polynomials that have a prescribed multiple eigenvalue.

Consider a matrix polynomial (m.p.)

 $P(\lambda) = A_m \lambda^m + A_{m-1} \lambda^{m-1} + \dots + A_1 \lambda + A_0,$

where λ is a complex variable, $A_j \in \mathbb{C}^{n \times n}$, and $\det P(\lambda) \not\equiv 0$.

A scalar $\lambda_0 \in \mathbb{C}$ is an **eigenvalue** of $P(\lambda)$ if $P(\lambda_0)x_0 = 0$ for some $0 \neq x_0 \in \mathbb{C}^n$, known as a **(right) eigenvector** of $P(\lambda)$ corresponding to λ_0 . The set of all eigenvalues of $P(\lambda)$,

 $\sigma(P) = \{\lambda \in \mathbb{C} : \det P(\lambda) = 0\},\$

is the **spectrum** of $P(\lambda)$ and contains at most nm (finite) elements. For a $\lambda_0 \in \sigma(P)$, its **algebraic multiplicity (a.m.)** is the multiplicity of λ_0 as a zero of det $P(\lambda)$, and its **geometric multiplicity (g.m.)** is the dimension of the null space of $P(\lambda_0)$. It holds that $a.m. \geq g.m$.

Suppose that for a $\lambda_0 \in \sigma(P)$, there are $x_0 \neq 0$, $x_1, \ldots, x_k \in \mathbb{C}^n$, s.t.

$$\sum_{j=0}^{\zeta} \frac{1}{j!} P^{(j)}(\lambda_0) x_{\xi-j} = 0 ; \quad \xi = 0, 1, \dots, k.$$

Then x_0 is an eigenvector of λ_0 , and x_1, x_2, \ldots, x_k are known as generalized eigenvectors. The set $\{x_0, x_1, \ldots, x_k\}$ is said to be a Jordan chain of $P(\lambda)$ corresponding to $\lambda_0 \in \sigma(P)$.

Any eigenvalue of $P(\lambda)$ of g.m. p has p maximal Jordan chains associated with p linearly independent eigenvectors, with total number of vectors equal to its a.m. The largest length of Jordan chains of $P(\lambda)$ corresponding to a $\lambda_0 \in \sigma(P)$ is called the **index** of λ_0 , and it is the size of the largest Jordan blocks of the Jordan canonical form of $P(\lambda)$ associated with λ_0 . The index is equal to 1 **iff** a.m. = g.m.

We are interested in the spectra of perturbations of $P(\lambda)$ of the form

$$Q(\lambda) = P(\lambda) + \Delta(\lambda) = \sum_{j=0}^{m} (A_j + \Delta_j) \lambda^j$$

for arbitrary $\Delta_j \in \mathbb{C}^{n \times n}$. For $w = \{w_0, w_1, \dots, w_m\}, w_j \ge 0 \ (j > 0), w_0 > 0$, and $\varepsilon > 0$, we define the set of perturbations of $P(\lambda)$,

$$\mathcal{B}(P,\varepsilon,\mathsf{w}) = \left\{ Q(\lambda) = \sum_{j=0}^{m} (A_j + \Delta_j)\lambda^j : \|\Delta_j\|_2 \le \varepsilon w_j, \ j = 0, 1, \dots, m \right\}.$$

The weights w_i allow freedom in how perturbations are measured.

The ε -pseudospectrum of $P(\lambda)$ [Tisseur-Higham, 2001] is defined as

 $\sigma_{\varepsilon}(P) = \left\{ \mu \in \sigma(Q) : Q(\lambda) \in \mathcal{B}(P, \varepsilon, \mathbf{w}) \right\},\$

and allows a "visualization" of the sensitivity of eigenvalues.

For the m.p. $P(\lambda)$ and a given $\mu \in \mathbb{C}$, we define the **distance from** $P(\lambda)$ to μ as an eigenvalue of g.m. at least r,

 $\mathcal{G}_r(\mu) = \inf \left\{ \varepsilon \ge 0 : \exists Q(\lambda) \in \mathcal{B}(P, \varepsilon, \mathsf{w}) \text{ with } \mu \in \sigma(Q) \text{ of g.m.} \ge r \right\},\$

the distance from $P(\lambda)$ to μ as an eigenvalue of a.m. at least r,

 $\mathcal{E}_r(\mu) = \inf \left\{ \varepsilon \ge 0 : \exists Q(\lambda) \in \mathcal{B}(P, \varepsilon, \mathsf{w}) \text{ with } \mu \in \sigma(Q) \text{ of a.m.} \ge r \right\},\$

and the distance from $P(\lambda)$ to μ as an eigenvalue of a.m. at least r and index (exactly) k,

 $\mathcal{E}_{r,k}(\mu) = \inf \{ \varepsilon \ge 0 : \exists Q(\lambda) \in \mathcal{B}(P, \varepsilon, w) \text{ with } \mu \in \sigma(Q) \}$

of a.m. $\geq r$ and index k.

Computation of $\mathcal{G}_r(\mu)$

(generalizing [Schmidt, 1907] & [Eckart-Young, 1936])

For a given $\mu \in \mathbb{C}$, consider the **SVD** of the (constant) matrix $P(\mu)$

 $P(\mu) = U \operatorname{diag} \{ s_1(P(\mu)), s_2(P(\mu)), \dots, s_n(P(\mu)) \} V^*,$

where $U = [u_1 \ u_2 \ \cdots \ u_n], V = [v_1 \ v_2 \ \cdots \ v_n] \in \mathbb{C}^{n \times n}$ are unitary. Then define the $n \times n$ matrices

$$E = -U \operatorname{diag}\{0, \dots, 0, s_{n-r+1}(P(\mu)), \dots, s_n(P(\mu))\} V^*$$

and

$$\Delta_j = \frac{w_j}{w(|\mu|)} \left(\frac{\overline{\mu}}{|\mu|}\right)^j E ; \quad j = 0, 1, \dots, m,$$

assuming that $\overline{\mu}/|\mu| = 0$ whenever $\mu = 0$.

The m.p.
$$\Delta(\lambda) = \sum_{j=0}^{m} \Delta_j \lambda^j$$
 satisfies $\Delta(\mu) = E$, and the perturbation
 $Q(\lambda) = P(\lambda) + \Delta(\lambda) = \sum_{j=0}^{m} (A_j + \Delta_j)\lambda^j$
lies on $\partial \mathcal{B}\left(P, \frac{s_{n-r+1}(P(\mu))}{w(|\mu|)}, w\right)$. Moreover, for every $j = n - r + 1$,
 $\dots, n - 1, n$, we have $Q(\mu)v_j = 0$ and $u_j^*Q(\mu) = 0$.

Theorem 1 The distance from $P(\lambda)$ to μ as an eigenvalue of geometric multiplicity r is

$$\mathcal{G}_r(\mu) = \frac{s_{n-r+1}(P(\mu))}{w(|\mu|)},$$

and $Q(\lambda)$ above is an optimal perturbation.

A lower bound for $\mathcal{E}_{r,k}(\mu)$ (generalizing [Malyshev, 1999])

By our definitions and the previous theorem,

$$\mathcal{E}_{1}(\mu) = \frac{s_{n}(P(\mu))}{w(|\mu|)} \leq \mathcal{E}_{r}(\mu) \leq \frac{s_{n-r+1}(P(\mu))}{w(|\mu|)} = \mathcal{G}_{r}(\mu) \leq \mathcal{E}_{r,1}(\mu).$$

For $0 \neq \gamma \in \mathbb{C}$, we define
$$F_{k}[P(\lambda);\gamma] = \begin{bmatrix} P(\lambda) & 0 & \cdots & 0\\ \gamma P^{(1)}(\lambda) & P(\lambda) & \cdots & 0\\ \frac{\gamma^{2}}{2!} P^{(2)}(\lambda) & \gamma P^{(1)}(\lambda) & \cdots & 0\\ \vdots & \vdots & \ddots & \vdots\\ \frac{\gamma^{k-1}}{(k-1)!} P^{(k-1)}(\lambda) & \frac{\gamma^{k-2}}{(k-2)!} P^{(k-2)}(\lambda) & \cdots & P(\lambda) \end{bmatrix}.$$

Lemma 2 Suppose $\lambda_0 \in \mathbb{C}$ is an eigenvalue of $P(\lambda)$ of a.m. at least r and index k. Then for any scalar $\gamma \neq 0$,

 $s_{kn-r+1}(F_k[P(\lambda_0);\gamma]) = 0.$

Theorem 3 For any $\gamma > 0$,

$$\mathcal{E}_{r,k}(\mu) \ge \frac{s_{kn-r+1}(F_k[P(\mu);\gamma])}{\|F_k[w(|\mu|);\gamma]\|}; \ k = 1, 2, \dots, r$$

and

$$\mathcal{E}_{r}(\mu) \geq \min_{k=1,2,\dots,nm} \frac{s_{kn-r+1}(F_{k}[P(\mu);\gamma])}{\|F_{k}[w(|\mu|);\gamma]\|}$$

An upper bound for $\mathcal{E}_r(\mu)$ (generalizing [Malyshev, 1999])

Without loss of generality, we may assume that $\gamma > 0$.

For
$$r \in \{2, 3, ..., n\}$$
, let $\begin{bmatrix} u_1(\gamma) \\ u_2(\gamma) \\ \vdots \\ u_r(\gamma) \end{bmatrix}$, $\begin{bmatrix} v_1(\gamma) \\ v_2(\gamma) \\ \vdots \\ v_r(\gamma) \end{bmatrix} \in \mathbb{C}^{rn}$, with

 $u_j(\gamma), v_j(\gamma) \in \mathbb{C}^n \ (j = 1, 2, ..., r)$, be left and right singular vectors of $s_{rn-r+1}(F_r[P(\mu); \gamma])$, respectively. We define the $n \times r$ matrices $U(\gamma) = [u_1(\gamma) \ u_2(\gamma) \ \cdots \ u_r(\gamma)]$ and $V(\gamma) = [v_1(\gamma) \ v_2(\gamma) \ \cdots \ v_r(\gamma)]$. If $rank(V(\gamma)) = r$, then we can construct a m.p. $\Delta_{\gamma}(\lambda)$ s.t. $Q_{\gamma}(\lambda) = P(\lambda) + \Delta_{\gamma}(\lambda)$ has μ as an eigenvalue of a.m. at least r.

We define the quantities $\phi_i = \frac{w^{(i)}(|\mu|)}{(i!) w(|\mu|)} \left(\frac{\overline{\mu}}{|\mu|}\right)^i$, i = 1, 2, ..., r, setting $\overline{\mu}/|\mu| = 0$ whenever $\mu = 0$. Then we define the $r \times r$ upper triangular Toeplitz matrix

$$\Theta_{\gamma} = [\theta_{i,j}] = \begin{bmatrix} 1 & -\gamma\phi_1 & \gamma^2(\phi_1^2 - \phi_2) & \gamma^3(2\phi_1\phi_2 - \phi_3 - \phi_1^3) & \cdots \\ 0 & 1 & -\gamma\phi_1 & \gamma^2(\phi_1^2 - \phi_2) & \cdots \\ 0 & 0 & 1 & -\gamma\phi_1 & \cdots \\ 0 & 0 & 0 & 1 & \cdots \\ \vdots & \vdots & \vdots & \ddots \end{bmatrix},$$

whose entries above the diagonal are given by the recursive formulae

$$\theta_{i,j} = -\theta_{i,i}\gamma^{j-i}\phi_{j-i} - \theta_{i,i+1}\gamma^{j-(i+1)}\phi_{j-(i+1)} - \dots - \theta_{i,j-1}\gamma\phi_1; \ 1 \le i < j \le r.$$

Denoting by $V(\gamma)^{\dagger}$ the Moore-Penrose pseudoinverse of $V(\gamma)$, we consider the $n \times n$ matrix

$$\Delta_{\gamma} = -s_{rn-r+1}(F_r[P(\mu);\gamma]) U(\gamma) \Theta_{\gamma} V(\gamma)^{\dagger},$$

and we define the matrices

$$\Delta_{\gamma,j} = \frac{w_j}{w(|\mu|)} \left(\frac{\overline{\mu}}{|\mu|}\right)^j \Delta_{\gamma}; \quad j = 0, 1, \dots, m$$

and the matrix polynomial

$$\Delta_{\gamma}(\lambda) = \sum_{j=0}^{m} \Delta_{\gamma,j} \lambda^{j}.$$

Then, the scalar μ is a multiple eigenvalue of the m.p.

$$Q_{\gamma}(\lambda) = P(\lambda) + \Delta_{\gamma}(\lambda) = \sum_{j=0}^{m} (A_j + \Delta_{\gamma,j}) \lambda^j$$

with $\{v_1(\gamma), \gamma^{-1}v_2(\gamma), \gamma^{-2}v_3(\gamma), \dots, \gamma^{-(r-1)}v_r(\gamma)\}\$ as an associated Jordan chain of length r.

Theorem 4 For any real $\gamma > 0$ s.t. $rank(V(\gamma)) = r \le n$, it holds that

$$\mathcal{E}_{r}(\mu) \leq \frac{s_{rn-r+1}(F_{r}[P(\mu);\gamma])}{w(|\mu|)} \left\| U(\gamma) \Theta_{\gamma} V(\gamma)^{\dagger} \right\|,$$

and $Q_{\gamma}(\lambda) \in \partial \mathcal{B}\left(P, \frac{s_{rn-r+1}(F_r[P(\mu);\gamma])}{w(|\mu|)} \| U(\gamma) \Theta_{\gamma} V(\gamma)^{\dagger} \|, w\right)$ above has μ as an eigenvalue with a Jordan chain of length at least r.

A numerical example

Let
$$P(\lambda) = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 3 \end{bmatrix} \lambda^2 + \begin{bmatrix} 0 & 1 & 0 \\ 0 & 3 & 1 \\ 0 & -1 & 6 \end{bmatrix} \lambda + \begin{bmatrix} 2 & 1 & 0 \\ -1 & 3 & 0 \\ 0 & 0 & 10 \end{bmatrix}$$
 and
w = {10, 6.11, 3}. The ε -pseudospectra of $P(\lambda)$ for $\varepsilon = 0.05, 0.1002, 0.16$, are:



For the scalar $\mu = 3 + i$, we get

$$\mathcal{G}_2(3+i) = \frac{s_2(P(3+i))}{w(|3+i|)} = \frac{32.1524}{59.3240} = 0.5420,$$

with an optimal perturbation

$$\begin{split} \hat{Q}(\lambda) &= \begin{bmatrix} 0.4134 + i0.0607 & -0.1918 + i0.0808 & -0.0090 - i0.0002 \\ 0.0403 - i0.0302 & 0.4100 + i0.2467 & -0.0611 - i0.0173 \\ -0.0037 + i0.0010 & 0.0599 + i0.0176 & 3.0019 + i0.0016 \end{bmatrix} \lambda^2 \\ &+ \begin{bmatrix} -1.1726 - i0.2606 & 0.5773 + i0.0325 & -0.0172 - i0.0062 \\ 0.0974 - i0.0325 & -0.2315 - i0.5476 & 0.8931 - i0.0727 \\ -0.0077 - i0.0004 & -0.8955 + i0.0727 & 6.0025 + i0.0043 \end{bmatrix} \lambda \\ &+ \begin{bmatrix} 0.3144 - i1.0114 & 0.3269 - i0.1683 & -0.0235 - i0.0185 \\ -0.8319 + i0.0000 & -1.7334 - i2.5223 & -0.1283 - i0.1682 \\ -0.0118 - i0.0046 & 0.1245 + i0.1669 & 10.0017 + i0.0080 \end{bmatrix} \end{split}$$

that has $\mu = 3 + i$, as an eigenvalue with a.m. = g.m. = 2.

The graphs of the lower bound for $\mathcal{E}_{2,2}(3+i)$ and the upper bound for $\mathcal{E}_2(3+i)$ are illustrated below, and for $\gamma = 1.9$,

 $0.2149 \leq \mathcal{E}_{2,2}(3+i) = \mathcal{E}_2(3+i) \leq 0.4901 < 0.5420 = \mathcal{G}_2(3+i).$



For $\mu = -1.1105$, a pseudospectra approach [Boulton-Lancaster-Ps., 2008] implies that

 $\mathcal{E}_2(-1.1105) = \mathcal{E}_{2,2}(-1.1105) = \mathcal{E}_1(-1.1105) = 0.1002.$

The graphs of the lower bound for $\mathcal{E}_{3,3}(-1.1105)$ and the upper bound for $\mathcal{E}_3(-1.1105)$ are given below. For $\gamma = 0.5530$ and $\gamma = 0.6518$, we get, respectively,

 $\mathcal{E}_{3,3}(-1.1105) \geq 0.1048$ and $\mathcal{E}_{3}(-1.1105) \leq 0.3177$.



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