

Nearest Pencils with Specified Eigenvalues

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Outline

- 1 Introduction
 - Kronecker Canonical Form
 - Problem Definition
 - Motivation
- 2 Derivation of a Singular Value Characterization
 - Rank Characterization
 - Construction of an Optimal Perturbation
 - Simultaneous Perturbations
- 3 Numerical Examples

Kronecker Canonical Form (KCF)

Given a pencil $(A - \lambda B)$ where $A, B \in \mathbb{C}^{m \times n}$ with $m \geq n$.

- There exist invertible matrices $P \in \mathbb{C}^{m \times m}$ and $Q \in \mathbb{C}^{n \times n}$ such that

$$P(A - \lambda B)Q = \text{diag}(A_1 - \lambda B_1, A_2 - \lambda B_2, \dots, A_k - \lambda B_k)$$

Each $A_j - \lambda B_j$ is in one of the following forms.

- (i) Jordan Block (associated with a finite eigenvalue α)

$$A_j - \lambda B_j = \begin{bmatrix} \lambda - \alpha & 1 & 0 & \dots & 0 \\ 0 & \lambda - \alpha & 1 & \dots & 0 \\ & & & \lambda - \alpha & 1 \\ 0 & & & 0 & \lambda - \alpha \end{bmatrix}$$

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Kronecker Canonical Form (KCF)

(ii) Block associated with an infinite eigenvalue

$$A_j - \lambda B_j = \begin{bmatrix} 1 & \lambda & 0 & \dots & 0 \\ 0 & 1 & \lambda & \dots & 0 \\ & & & 1 & \lambda \\ 0 & & & 0 & 1 \end{bmatrix}$$

(iii) Singular blocks of the form (of size $(n_j + 1) \times n_j$)

$$A_j - \lambda B_j = \begin{bmatrix} \lambda & 0 & & & 0 \\ 1 & \lambda & & & 0 \\ 0 & 1 & \lambda & & \\ & & & & \\ & & & & 1 & \lambda \\ 0 & & & & 0 & 1 \end{bmatrix}$$

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 - Simultaneous Perturbations
- 3 **Numerical Examples**

Problem Definition

- Algebraic multiplicity of an eigenvalue α : Sum of the sizes of the Jordan blocks associated with α in the KCF.
- $\lambda_1, \dots, \lambda_k \in \mathbb{C}$ be given scalars, and $\mathcal{S} = \{\lambda_1, \lambda_2, \dots, \lambda_k\}$
- r be a given positive integer
- $m_j(A, B)$: Algebraic multiplicity of λ_j as an eigenvalue of $A - \lambda B$

Definition (Distance to Pencils with Specified Eigenvalues)

$$\tau_r(A, B, \mathcal{S}) = \inf \left\{ \left\| \begin{bmatrix} \delta A & \delta B \end{bmatrix} \right\|_2 : \sum_{j=1}^k m_j(A + \alpha_A \delta A, B + \alpha_B \delta B) \geq r \right\}$$

$$\alpha_A = 1, \alpha_B = 0 \quad \text{or} \quad \alpha_A = \alpha_B = 1$$

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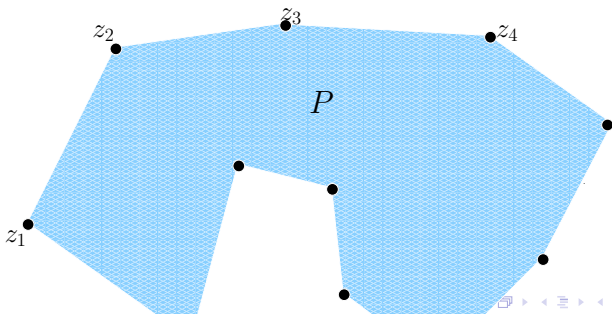
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Shape Estimation from Moments

- The distance to the nearest matrix with a multiple eigenvalue is a special case.

$$\inf_{\lambda \in \mathbb{C}} \tau_2(A, I, \lambda)$$

- Estimating a polygon from moments (Elad, Milanfar, Golub; 2004)



Shape Estimation from Moments

- Given moments

$$\mathcal{M}_k = \int \int_P z^k dx dy$$

for $k = 1, \dots, m$.

Estimate the vertices $z_j \in \mathbb{C}$ for $j = 1, \dots, n$ of P .

- The vertices z_j are the eigenvalues of a pencil $T_0 - \lambda T_1$ where $T_0, T_1 \in \mathbb{C}^{m \times n}$ (with $m \geq n$) are Hankel matrices defined in terms of \mathcal{M}_k .
- Because of measurement errors the perturbed pencil $\tilde{T}_0 - \lambda \tilde{T}_1$ has generically no eigenvalues.
- Find a nearest pencil with a full set of eigenvalues

$$\inf_{S \in \mathbb{C}^n} \tau_n(\tilde{T}_0, \tilde{T}_1, S).$$

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Derivation of a Singular Value Characterization

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Rank Characterization

- Suppose $\alpha_A = 1$ and $\alpha_B = 0$.
- First step in derivation is a rank characterization for $\sum_{j=1}^k m_j(A, B) \geq r$.
- We benefit from a Sylvester operator point of view.

Theorem (Kernel of Sylvester Operator)

Let $A, B \in \mathbb{C}^{m \times n}$ be such that $m \geq n$ and $\text{rank}(B) = n$, and $C \in \mathbb{C}^{r \times r}$. The solution space of the generalized Sylvester equation

$$AX - BXC = 0$$

depends on the Kronecker canonical form of $A - \lambda B$ and the Jordan form of C . Specifically suppose that μ_1, \dots, μ_ℓ are the common eigenvalues of $A - \lambda B$ and C . Let $c_{j,1}, \dots, c_{j,\ell_j}$ and $p_{j,1}, \dots, p_{j,\tilde{\ell}_j}$ denote the sizes of the Jordan blocks of $A - \lambda B$ and C associated with the eigenvalue μ_j , respectively. Then

$$\dim\{X \in \mathbb{C}^{n \times r} : AX - BXC = 0\} = \sum_{j=1}^{\ell} \sum_{i=1}^{\ell_j} \sum_{q=1}^{\tilde{\ell}_j} \min(c_{j,i}, p_{j,q}).$$

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Rank Characterization

- Suppose $C = C(\mu, \Gamma) = \begin{bmatrix} \mu_1 & -\gamma_{21} & \dots & -\gamma_{r1} \\ 0 & \mu_2 & \dots & -\gamma_{r2} \\ & & \ddots & \\ 0 & & & \mu_r \end{bmatrix}$, and
 $\mathcal{G} = \{\Gamma : C(\mu, \Gamma) \text{ has Jordan blocks of maximal size.}\}$

Theorem (Sylvester Characterization)

Given a pencil $A - \lambda B$ with $A, B \in \mathbb{C}^{m \times n}$ such that $m \geq n$ and $\text{rank}(B) = n$, a set $S = \{\lambda_1, \dots, \lambda_k\}$ consisting of distinct complex scalars and a positive integer r . Then the following two statements are equivalent.

- $\sum_{j=1}^k m_j(A, B) \geq r$
- There exists a $\mu \in S^r$ such that

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for all $\Gamma \in \mathcal{G}$.

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Rank Characterization

Kroneckerization of the Sylvester Equation

- Recall the basic Identity for $X = [X_1 \quad \dots \quad X_r] \in \mathbb{C}^{n \times r}$

$$\text{vec}(FXG) = (G^T \otimes F)\text{vec}(X), \quad \text{where } \text{vec}(X) = \begin{bmatrix} X_1 \\ X_2 \\ \vdots \\ X_r \end{bmatrix} \in \mathbb{C}^{nr}$$

- In particular

$$AX - BXC(\mu, \Gamma) = 0 \Leftrightarrow ((I \otimes A) - (C^T(\mu, \Gamma) \otimes B))\text{vec}(X) = 0.$$

- Consequently

$$\dim\{X \in \mathbb{C}^{n \times r} : AX - BXC(\mu, \Gamma) = 0\} \geq r$$

$$\iff \text{rank}(((I \otimes A) - (C^T(\mu, \Gamma) \otimes B))) \leq nr - r$$

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$$\iff \text{rank}(((I \otimes A) - (C^T(\mu, \Gamma) \otimes B))) \leq nr - r$$

Rank Characterization

Kroneckerization of the Sylvester Equation

- Recall the basic Identity for $X = [X_1 \quad \dots \quad X_r] \in \mathbb{C}^{n \times r}$

$$\text{vec}(FXG) = (G^T \otimes F)\text{vec}(X), \quad \text{where } \text{vec}(X) = \begin{bmatrix} X_1 \\ X_2 \\ \vdots \\ X_r \end{bmatrix} \in \mathbb{C}^{nr}$$

- In particular
 $AX - BXC(\mu, \Gamma) = 0 \Leftrightarrow ((I \otimes A) - (C^T(\mu, \Gamma) \otimes B))\text{vec}(X) = 0.$

- Consequently

$$\dim\{X \in \mathbb{C}^{n \times r} : AX - BXC(\mu, \Gamma) = 0\} \geq r$$

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Rank Characterization

Kroneckerization of the Sylvester Equation

$$\begin{aligned} \mathcal{L}(\mu, \Gamma, A, B) &:= \left((I \otimes A) - (C^T(\mu, \Gamma) \otimes B) \right) \\ &= \begin{bmatrix} A - \mu_1 B & 0 & & & 0 \\ \gamma_{21} B & A - \mu_2 B & & & 0 \\ & & \ddots & & \\ \gamma_{r1} B & \gamma_{r2} B & & A - \mu_{r-1} B & 0 \\ & & & \gamma_{r(r-1)} B & A - \mu_r B \end{bmatrix}. \end{aligned}$$

Theorem (Rank Characterization)

Given a pencil $A - \lambda B$ with $A, B \in \mathbb{C}^{n \times m}$ such that $n \geq m$, a set $S = \{\lambda_1, \dots, \lambda_k\}$ consisting of distinct complex scalars and a positive integer r . Then the following two statements are equivalent.

- 1 $\sum_{j=1}^k m_j(A, B) \geq r$
- 2 There exists a $\mu \in S^r$ such that

$$\text{rank}(\mathcal{L}(\mu, \Gamma, A, B)) \leq nr - r$$

for all $\Gamma \in \mathcal{G}$.

Rank Characterization

The rank characterization yields a lower bound on the distance.

Theorem (Minimal Rank)

Given $C \in \mathbb{C}^{\ell \times q}$ and a positive integer $p < \min(\ell, q)$. Then

$$\inf\{\|\delta C\|_2 : \text{rank}(C + \delta C) \leq p\} = \sigma_{p+1}(C).$$

- From an application of the theorem above

$$\begin{aligned} \tau_r(A, B, S) &= \inf_{\mu \in S^r} \inf\{\|\delta A\| : \forall \Gamma \in \mathcal{G} \text{ rank}(\mathcal{L}(\mu, \Gamma, A + \delta A, B)) \leq nr - r\} \\ &\geq \inf_{\mu \in S^r} \sup_{\Gamma} \sigma_{nr-r+1}(\mathcal{L}(\mu, \Gamma, A, B)) \end{aligned}$$

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Outline

- 1 Introduction
 - Kronecker Canonical Form
 - Problem Definition
 - Motivation
- 2 Derivation of a Singular Value Characterization
 - Rank Characterization
 - Construction of an Optimal Perturbation
 - Simultaneous Perturbations
- 3 Numerical Examples

Construction of an Optimal Perturbation

We deduce the other direction (for all μ)

$$\inf\{\|\delta\mathbf{A}\| : \forall \Gamma \in \mathcal{G} \text{ rank}(\mathcal{L}(\mu, \Gamma, \mathbf{A} + \delta\mathbf{A}, \mathbf{B})) \leq nr - r\} \\ \leq \\ \sup_{\Gamma} \sigma_{nr-r+1}(\mathcal{L}(\mu, \Gamma, \mathbf{A}, \mathbf{B}))$$

under mild **multiplicity**, **linear independence**, and **full Jordan block** assumptions by constructing an optimal perturbation.

Construction of an Optimal Perturbation

Let $\kappa_r(\mu) := \sup_{\Gamma} \sigma_{nr-r+1}(\mathcal{L}(\mu, \Gamma, A, B))$.

Construct a perturbation δA_* such that

- 1 $\|\delta A_*\| = \kappa_r(\mu)$
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Construction of an Optimal Perturbation

Specifically $\delta A_* := -\kappa_r(\mu)\mathcal{U}\mathcal{V}^+$.

Above $\mathcal{U} \in \mathbb{C}^{m \times r}$ and $\mathcal{V} \in \mathbb{C}^{n \times r}$ are defined as follows.

- Let Γ_* be the optimal Γ satisfying

$$\kappa_r(\mu) = \sigma_{nr-r+1}(\mathcal{L}(\mu, \Gamma_*, A, B))$$

- Consider the left and right singular vectors assoc with $\kappa_r(\mu)$

$$\mathcal{L}(\mu, \Gamma_*, A, B) V = \kappa_r(\mu) U \quad \text{and} \quad U^* \mathcal{L}(\mu, \Gamma_*, A, B) = V^* \kappa_r(\mu).$$

- Then \mathcal{U} and \mathcal{V} are such that

$$\text{vec}(\mathcal{U}) = U \quad \text{and} \quad \text{vec}(\mathcal{V}) = V.$$

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Construction of an Optimal Perturbation

Theorem (Rellich, 1937-1942)

Let $f : \mathbb{R} \rightarrow \mathbb{R}$ be defined as $f(\omega) = \sigma_j(\mathcal{A}(\omega))$ (σ_j denoting the j th largest singular value) where $\mathcal{A}(\omega)$ is an analytic matrix-valued function of ω . If the multiplicity of $\sigma_j(\mathcal{A}(\tilde{\omega}))$ is one and $\sigma_j(\mathcal{A}(\tilde{\omega})) \neq 0$, then $f(\omega)$ is real analytic at $\tilde{\omega}$ with the derivative

$$f'(\tilde{\omega}) = \text{Real} \left(u^* \frac{d\mathcal{A}(\tilde{\omega})}{d\omega} v \right)$$

where u and v consist of a consistent pair of a unit left and a right singular vector associated with $\sigma_j(\mathcal{A}(\tilde{\omega}))$.

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$$(1) \quad \|\delta A_*\| = \|- \kappa_r(\mu) \mathcal{U} \mathcal{V}^+\| = \kappa_r(\mu)$$

- Applying the theorem (Rellich) to $\sigma_{nr-r+1}(\mathcal{L}(\mu, \Gamma, A, B))$, the optimality of Γ_* implies

$$\mathcal{U}^* \mathcal{U} = \mathcal{V}^* \mathcal{V}$$

- Consider

$$\mathcal{U} \mathcal{V}^+ = \mathcal{U} \mathcal{V}^+ \mathcal{V} \mathcal{V}^+$$

- $\mathcal{V} \mathcal{V}^+$ first orthogonally projects a vector onto $\text{Range}(\mathcal{V})$
- $\mathcal{U} \mathcal{V}^+$ then changes the coordinates from \mathcal{V} to \mathcal{U}
- Consequently $\|\mathcal{U} \mathcal{V}^+\| = 1$.

Construction of an Optimal Perturbation

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$$(2) \quad \text{rank}(\mathcal{L}(\mu, \Gamma_*, A + \delta A_*, B)) \leq nr - r$$

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- Now any matrix in the subspace

$$\{\mathcal{V}D : C(\mu, \Gamma)D - DC(\mu, \Gamma) = 0\}$$

of dimension at least r is a solution for the Sylvester equation *i.e.*,

$$(A + \delta A_*)\mathcal{V}D - BVC(\mu, \Gamma_*)D = 0 \iff (A + \delta A_*)(\mathcal{V}D) - B(\mathcal{V}D)C(\mu, \Gamma_*) = 0$$

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$$(2) \quad \text{rank}(\mathcal{L}(\mu, \Gamma_*, A + \delta A_*, B)) \leq nr - r$$

- Suppose that the columns of \mathcal{V} are linearly independent.

$$\begin{aligned} \mathcal{L}(\mu, \Gamma_*, A, B) V = \kappa_r(\mu) U &\iff ((I \otimes A) - (C^T(\mu, \Gamma_*) \otimes B)) V = \kappa_r(\mu) U \\ &\iff AV - BVC(\mu, \Gamma_*) = \kappa_r(\mu)U \\ &\iff AV - BVC(\mu, \Gamma_*) = \kappa_r(\mu)U(\mathcal{V}^+\mathcal{V}) \\ &\implies (A - \kappa_r(\mu)U\mathcal{V}^+)\mathcal{V} - BVC(\mu, \Gamma_*) = 0 \\ &\implies (A + \delta A_*)\mathcal{V} - BVC(\mu, \Gamma_*) = 0. \end{aligned}$$

- Now any matrix in the subspace

$$\{\mathcal{V}D : C(\mu, \Gamma)D - DC(\mu, \Gamma) = 0\}$$

of dimension at least r is a solution for the Sylvester equation *i.e.*,

$$(A + \delta A_*)\mathcal{V}D - BVC(\mu, \Gamma_*)D = 0 \iff (A + \delta A_*)(\mathcal{V}D) - B(\mathcal{V}D)C(\mu, \Gamma_*) = 0$$

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Construction of an Optimal Perturbation

Theorem (Nearest Pencils with Specified Eigenvalues)

Let $A - \lambda B$ be an $m \times n$ pencil with $m \geq n$, r be a positive integer and $\mathcal{S} = \{\lambda_1, \dots, \lambda_k\}$ be a set of distinct complex scalars.

1 Then the equality

$$\tau_r(A, B, \mathcal{S}) = \inf_{\mu \in \mathcal{S}^r} \sup_{\Gamma} \sigma_{nr-r+1}(\mathcal{L}(\mu, \Gamma, A, B))$$

holds provided that the optimization problem on the right is attained at a (μ_*, Γ_*) where the *multiplicity*, *linear independence* and *full Jordan block* qualifications hold.

2 Furthermore a minimal perturbation δA_* such that $\sum_{j=1}^k m(A + \delta A_*, B) \geq r$ is given by

$$\delta A_* = -\tau_r(A, B, \mathcal{S}) \mathcal{U} \mathcal{V}^+$$

where the matrices \mathcal{U} and \mathcal{V} are formed by the left and right singular vectors at the optimal (μ_*, Γ_*) .

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Outline

- 1 Introduction
 - Kronecker Canonical Form
 - Problem Definition
 - Motivation
- 2 Derivation of a Singular Value Characterization
 - Rank Characterization
 - Construction of an Optimal Perturbation
 - Simultaneous Perturbations
- 3 Numerical Examples

Simultaneous Perturbations

When $\alpha_A = \alpha_B = 1$ and $\tau_r(A, B, S)$ defined w.r.t. the Frobenius norm

$$\tau_r(A, B, S) \geq \inf_{\mu \in S^r} \sup_{\Gamma} \frac{\sigma_{nr-r+1}(\mathcal{L}(\mu, \Gamma, A, B))}{\sqrt{1 + \|C(\mu, \Gamma)\|_2^2}}$$

Numerical Examples

(1) Distance from $A - \lambda B$ to pencils with multiple eigenvalues

$$\inf_{\lambda \in \mathbb{C}} \sup_{\gamma \in \mathbb{C}} \sigma_{2n-1} \left(\begin{bmatrix} A - \lambda B & 0 \\ \gamma B & A - \lambda B \end{bmatrix} \right)$$

e.g.

$$A - \lambda B = \begin{bmatrix} 2 & -1 & -1 \\ -1 & 2 & -1 \\ -1 & -1 & 2 \end{bmatrix} - \lambda \begin{bmatrix} -1 & 2 & 3 \\ 2 & -1 & 2 \\ 4 & 2 & -1 \end{bmatrix}$$

The nearest pencil

$$\begin{bmatrix} 1.91465 & -0.57896 & -1.21173 \\ -1.32160 & 1.93256 & -0.57897 \\ -0.72082 & -1.32160 & 1.91466 \end{bmatrix} - \lambda \begin{bmatrix} -1 & 2 & 3 \\ 2 & -1 & 2 \\ 4 & 2 & -1 \end{bmatrix}$$

at a distance of 0.59299 with multiple eigenvalue $\lambda_* = -0.85488$.

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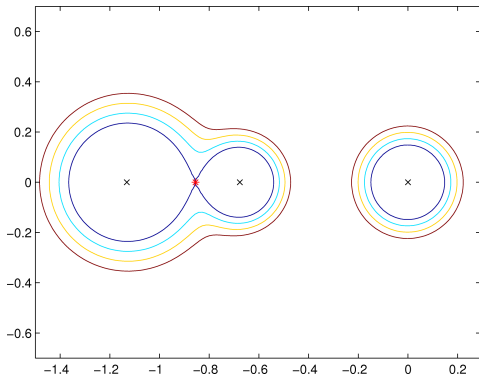
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The ϵ -pseudospectrum of $A - \lambda B$

$$\Lambda_\epsilon(A, B) = \{ \lambda \in \mathbb{C} : \exists (\delta A) \det(A + \delta A - \lambda B) = 0 \text{ and } \|\delta A\|_2 \leq \epsilon \}$$

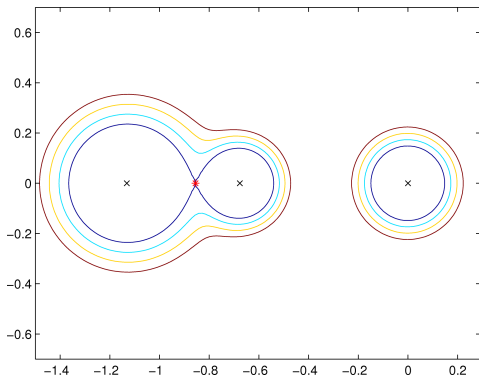


Blue Curve: $\Lambda_\epsilon(A, B)$ for $\epsilon = 0.59299$; Red asterisks marks $\lambda_* = -0.85488$.

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Numerical Examples

(ii) Distance from rectangular $A - \lambda B$ to pencils with r eigenvalues

$$\inf_{\lambda_j \in \mathbb{C}} \sup_{\gamma_{ik} \in \mathbb{C}} \sigma_{nr-r+1} \left(\begin{bmatrix} A - \lambda_1 B & 0 & & 0 \\ \gamma_{21} B & A - \lambda_2 B & & 0 \\ & & \ddots & \\ \gamma_{r1} B & \gamma_{r2} B & & A - \lambda_r B \end{bmatrix} \right)$$

e.g.

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at a distance of 0.03927 with two eigvals $\lambda_1^* = 2.55144$ and $\lambda_2^* = -1.45405$.

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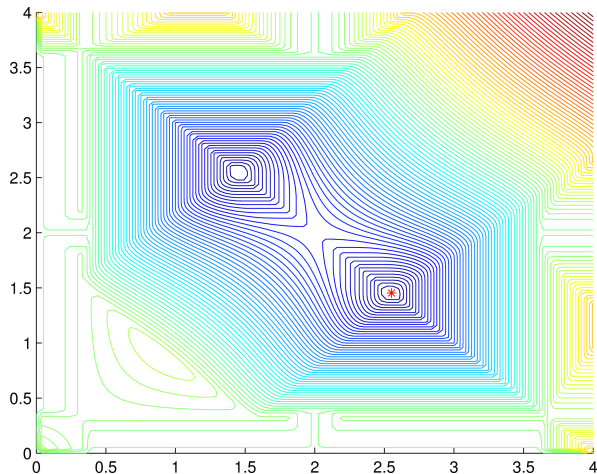
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Numerical Examples



Level sets of $g(\lambda_1, \lambda_2)$; **Red asterisks** marks $(\lambda_1^*, \lambda_2^*) = (2.5514, 1.4541)$.

Numerical Examples

(iii) Distance from $A - \lambda B$ to stable pencils

$$\inf_{\lambda_j \in \mathbb{C}^-} \sup_{\gamma_{p\ell} \in \mathbb{C}} \sigma_{nr-r+1} \left(\begin{bmatrix} A - \lambda_1 B & 0 & & 0 \\ \gamma_{21} B & A - \lambda_2 B & & 0 \\ & & \ddots & \\ \gamma_{m1} B & \gamma_{n2} B & & A - \lambda_n B \end{bmatrix} \right)$$

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$$A = \begin{bmatrix} 0.6 - \frac{i}{3} & -0.2 + \frac{4i}{3} \\ -0.1 + \frac{2i}{3} & 0.5 + \frac{i}{3} \end{bmatrix}$$

is unstable with eigenvalues $0.7 - i$ and $0.4 + i$ at a distance of 0.6610 to stability. The matrix

$$A + \delta A_* = \begin{bmatrix} 0.0681 - 0.3064i & -0.4629 + 1.2524i \\ 0.2047 + 0.5858i & -0.1573 + 0.3064i \end{bmatrix}$$

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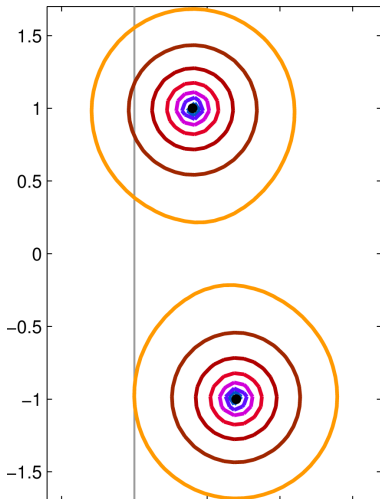
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Orange Curve:
 $\Lambda_\epsilon(A)$ for $\epsilon = 0.6610$

Summary

- A **singular value characterization** is derived for $\tau_r(A, B, \mathcal{S})$.
- Charac makes the **numerical computation** possible.

- Open Problems
 - A singular value charac for $\tau_r(A, B, \mathcal{S})$ when $\alpha_A = \alpha_B = 1$
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