

The Jacobian Method

The Art of Finding More Needles in Nearby Haystacks

A Ph.D. defense by
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University of Wyoming, July 2014

Introduction:

History, motivation
Definitions and preliminaries

Often for mathematicians finding a needle in a haystack can be formulated as solving

$$f(x, y) = c,$$

where $f : \mathbb{R}^{n+m} \rightarrow \mathbb{R}^n$



The Implicit Function Theorem says when a particular solution $f(x_0, y_0) = c$ is 'nice' then one can solve $f(x, y_1) = c$ for any y_1 near y_0 .

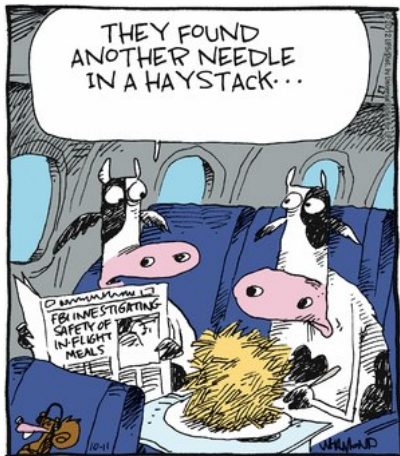
i.e. if you find a nice needle in the haystack $f(x, y_0) = c$,



then all nearby haystacks $f(x, y_1) = c$ have a needle.



Why do we care?



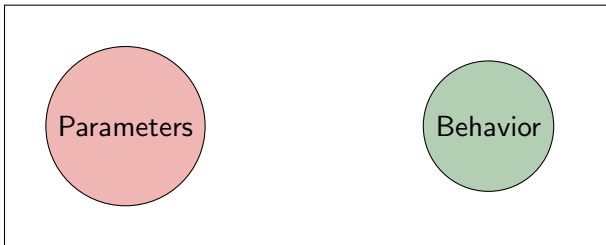


B. Parlett,
in *The Zahir*

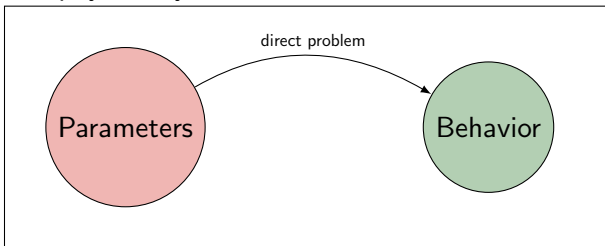
"Vibrations are everywhere, and so too are the eigenvalues associated with them."

Beresford N. Parlett, 1998

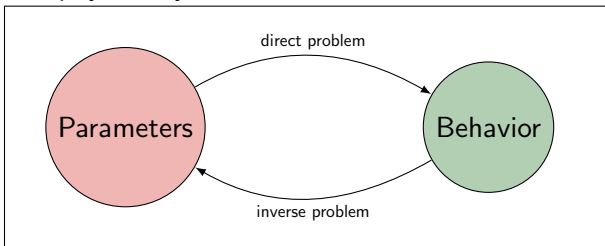
A physical system



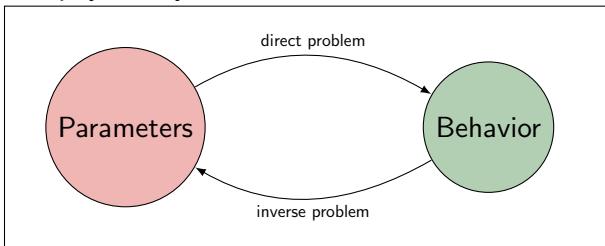
A physical system



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A physical system



- ▶ Agustin Cauchy (1789–1875)
- ▶ Jacques Sturm (1803–1855)
- ▶ Joseph Liouville (1809–1882)

A physical system:

A physical system:

Motion equations:

$$m_1 \ddot{x}_1 = -k_1 x_1 + k_2 (x_2 - x_1) + F_1(t),$$

$$m_2 \ddot{x}_2 = -k_3 x_2 - k_2 (x_2 - x_1) + F_2(t).$$

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Matrix form:

$$\begin{bmatrix} m_1 & \\ & m_2 \end{bmatrix} \begin{bmatrix} \ddot{x}_1 \\ \ddot{x}_2 \end{bmatrix} + \begin{bmatrix} k_1 + k_2 & -k_2 \\ -k_2 & k_2 + k_3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} F_1 \\ F_2 \end{bmatrix}.$$

$$\mathbf{A} \ddot{\mathbf{x}} + \mathbf{C} \mathbf{x} = \mathbf{F}$$

A physical system:

Motion equations:

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$$\mathbf{A} \ddot{\mathbf{x}} + \mathbf{C} \mathbf{x} = \mathbf{F}$$

When $\mathbf{F} = \mathbf{0}$, the eigenvalues of C describe the 'natural frequencies' of the system.

Graph of a matrix

$A_{n \times n}$: real symmetric matrix

$G(A)$: a graph G on n vertices $1, 2, \dots, n$

$i \sim j$ if and only if $i \neq j$ and $a_{ij} \neq 0$

$G(A)$ does not depend on the diagonal entries of A

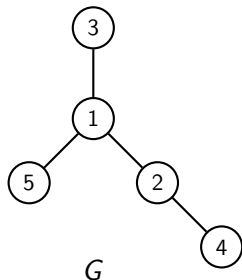
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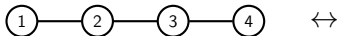
$G(A)$ does not depend on the diagonal entries of A



$$A = \begin{bmatrix} 2 & 1 & 3 & 0 & -4 \\ 1 & 0 & 0 & 1 & 0 \\ 3 & 0 & -1 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ -4 & 0 & 0 & 0 & 5 \end{bmatrix}$$

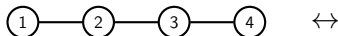
Then we say $A \in S(G)$.

- ▶ In general the coefficient matrix C for the system of masses and springs has a certain zero-nonzero pattern that can be described by its graph.



$$C = \begin{bmatrix} k_1 + k_2 & -k_2 & & \\ -k_2 & k_2 + k_3 & -k_3 & \\ & -k_3 & k_3 + k_4 & -k_4 \\ & & -k_4 & k_4 + k_5 \end{bmatrix}$$

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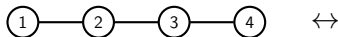


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Previous studies when graph of A is a:

- ▶ **star** [Fan, Pall 1957]

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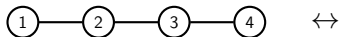


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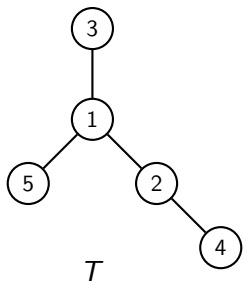


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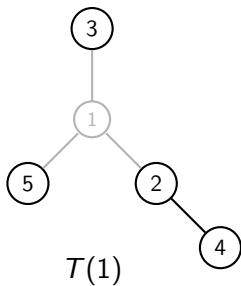
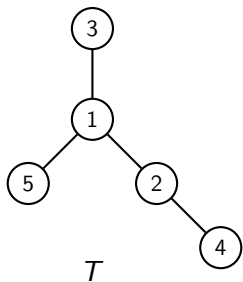
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- ▶ **star** [Fan, Pall 1957]
- ▶ **path** [Gladwell 1988]
- ▶ **tree** [Duarte 1989]

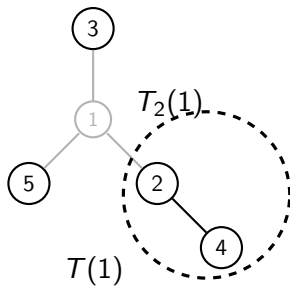
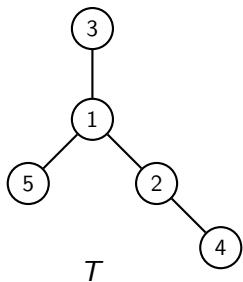
Submatrices and Subtrees



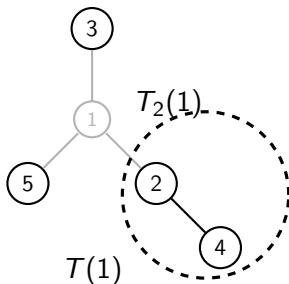
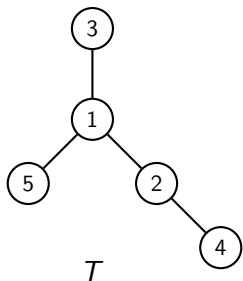
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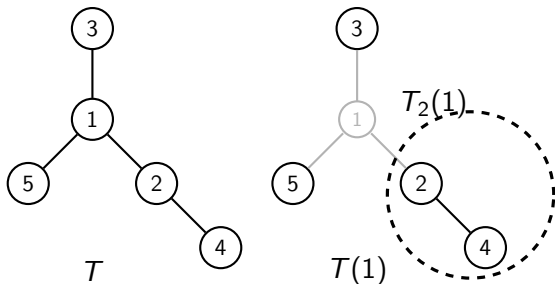


Submatrices and Subtrees



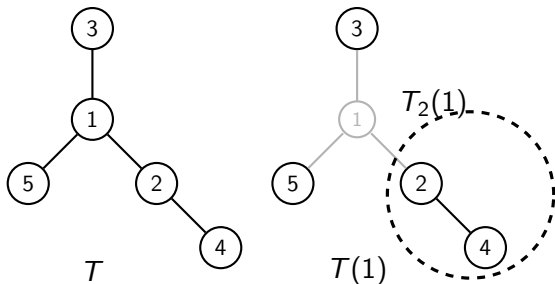
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Submatrices and Subtrees



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Submatrices and Subtrees



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Cauchy Interlacing Inequalities

$A_{n \times n}$: real symmetric matrix

Eigenvalues of A : $\lambda_1 \leq \lambda_2 \leq \cdots \leq \lambda_n$

Eigenvalues of $A(r)$: $\mu_1 \leq \mu_2 \leq \cdots \leq \mu_{n-1}$

Eigenvalues of $A(\{r, s\})$: $\tau_1 \leq \tau_2 \leq \cdots \leq \tau_{n-2}$

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Then

$$\lambda_i \leq \mu_i \leq \lambda_{i+1}, \quad i = 1, \dots, n-1,$$

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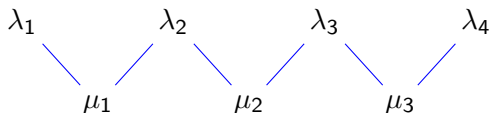
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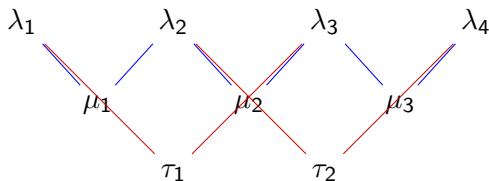
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The Implicit Function Theorem

Theorem

$$\mathbf{x} \in \mathbb{R}^s, \mathbf{y} \in \mathbb{R}^r$$

$F : \mathbb{R}^{s+r} \rightarrow \mathbb{R}^s$: continuously differentiable on an open subset U of \mathbb{R}^{s+r}

$$F(\mathbf{x}, \mathbf{y}) = (F_1(\mathbf{x}, \mathbf{y}), F_2(\mathbf{x}, \mathbf{y}), \dots, F_s(\mathbf{x}, \mathbf{y})),$$

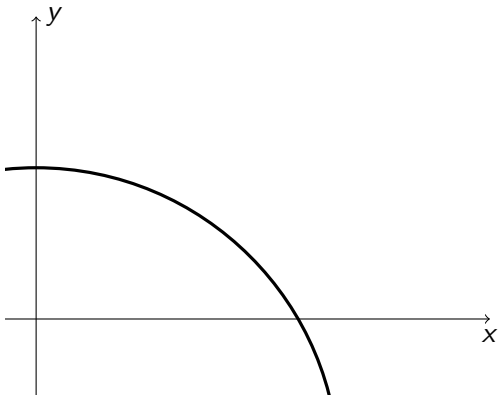
$(\mathbf{a}, \mathbf{b}) \in U$ with $\mathbf{a} \in \mathbb{R}^s$, $\mathbf{b} \in \mathbb{R}^r$

$\mathbf{c} \in \mathbb{R}^s$ such that $F(\mathbf{a}, \mathbf{b}) = \mathbf{c}$

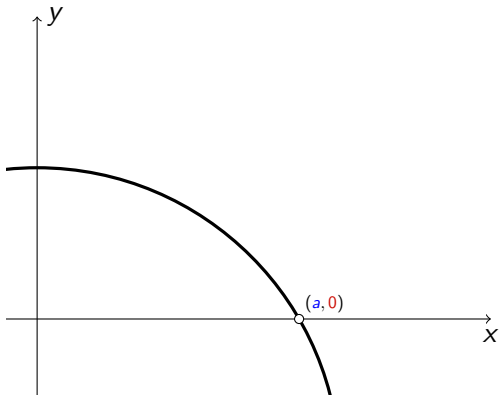
If $\left[\frac{\partial F_i}{\partial x_j} \Big|_{(\mathbf{a}, \mathbf{b})} \right]$ is nonsingular, then there exist an open neighborhood V containing \mathbf{a} and an open neighborhood W containing \mathbf{b} such that $V \times W \subseteq U$ and for each $\mathbf{y} \in W$ there is an $\mathbf{x} \in V$ with

$$F(\mathbf{x}, \mathbf{y}) = \mathbf{c}$$

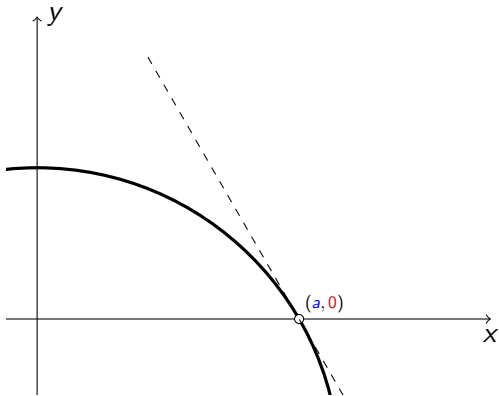
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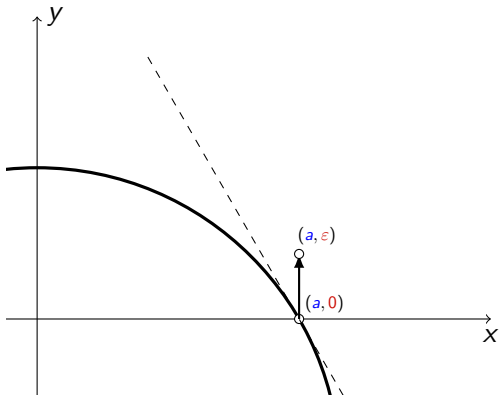
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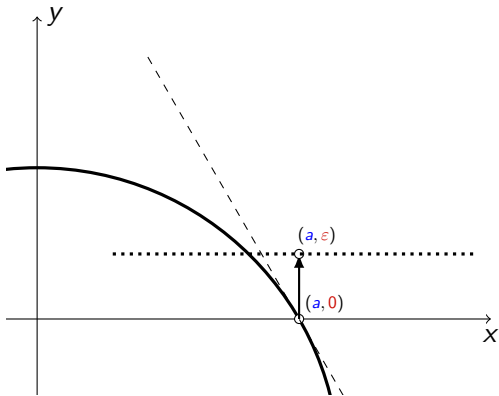
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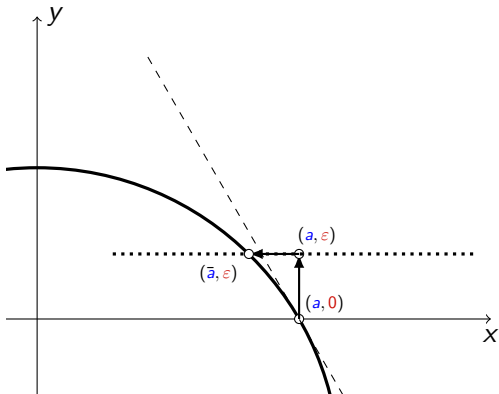
The Implicit Function Theorem



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The **Jacobian** Method

The λ -Structured Inverse Eigenvalue Problem

Theorem (K. H.M. and B.L. Shader 2014)

For given

$\lambda_1 < \lambda_2 < \dots < \lambda_n$: real numbers, and

G : a *graph* on n vertices,

there is an $n \times n$ real symmetric matrix A such that
eigenvalues of A are $\lambda_1, \lambda_2, \dots, \lambda_n$, and
graph of A is G .

The λ -Structured Inverse Eigenvalue Problem

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Proof:

Let

$$A = \begin{bmatrix} \lambda_1 & & & \\ & \lambda_2 & & \\ & & \dots & \\ & & & \lambda_n \end{bmatrix}$$

Define:

$$M = M(\mathbf{x}, \mathbf{y}) = \begin{bmatrix} x_1 & y_1 & \cdots & \\ y_1 & x_2 & \ddots & \vdots \\ \vdots & \ddots & \ddots & y_m \\ & \cdots & y_m & x_n \end{bmatrix}$$

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Note that $M(\lambda_1, \dots, \lambda_n, 0, \dots, 0) = A$.

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Note that $M(\lambda_1, \dots, \lambda_n, 0, \dots, 0) = A$.

Define:

$$F : (\mathbf{x}, \mathbf{y}) \mapsto (\lambda_1(M), \lambda_2(M), \dots, \lambda_n(M)).$$

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Define:

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Note that $F|_A = F(\lambda_1, \dots, \lambda_n, 0, \dots, 0) = (\lambda_1, \lambda_2, \dots, \lambda_n)$.

$$F : (\mathbf{x}, \mathbf{y}) \mapsto (\lambda_1(M), \lambda_2(M), \dots, \lambda_n(M)).$$

Then

$$\text{Jac}(F)|_A = \left[\begin{array}{c|c} I & O \end{array} \right]$$

has full row-rank.

$$F(\lambda_1, \dots, \lambda_n, 0, \dots, 0) = (\lambda_1, \lambda_2, \dots, \lambda_n).$$

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$$F(\lambda_1, \dots, \lambda_n, 0, \dots, 0) = (\lambda_1, \lambda_2, \dots, \lambda_n).$$

Then by the Implicit Function Theorem, for $\bar{\mathbf{y}} = (\varepsilon_1, \dots, \varepsilon_m)$ with sufficiently small $\varepsilon_i > 0$, there is an $\bar{\mathbf{x}} = (\bar{\lambda}_1, \dots, \bar{\lambda}_n)$ close to $(\lambda_1, \dots, \lambda_n)$ such that

$$F(\bar{\lambda}_1, \dots, \bar{\lambda}_n, \varepsilon_1, \dots, \varepsilon_m) = (\lambda_1, \lambda_2, \dots, \lambda_n).$$

$$F : (\mathbf{x}, \mathbf{y}) \mapsto (\lambda_1(M), \lambda_2(M), \dots, \lambda_n(M)).$$

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Let $\bar{A} = M(\bar{\mathbf{x}}, \bar{\mathbf{y}})$. Then graph of \bar{A} is G , and eigenvalues of \bar{A} are $\lambda_1, \dots, \lambda_n$. □

Three
Fundamental
Structured
Inverse Eigenvalue
Problems

The λ - μ -SIEP: for trees

Theorem (A. Leal-Duarte 1989)

For given

$\lambda_1 < \mu_1 < \lambda_2 < \mu_2 < \dots < \mu_{n-1} < \lambda_n$: real numbers,

G : a *tree* on n vertices, and

v : a fixed vertex of G ,

there is an $n \times n$ real symmetric matrix A such that

eigenvalues of A are $\lambda_1, \lambda_2, \dots, \lambda_n$,

eigenvalues of $A(v)$ are $\mu_1, \mu_2, \dots, \mu_{n-1}$, and

graph of A is G .

The λ - μ -SIEP: for connected graphs

Theorem (K. H.M. and B.L. Shader 2013)

For given

$\lambda_1 < \mu_1 < \lambda_2 < \mu_2 < \dots < \mu_{n-1} < \lambda_n$: real numbers,

G : a *connected graph* on n vertices, and

v : a fixed vertex of G ,

there is an $n \times n$ real symmetric matrix A such that

eigenvalues of A are $\lambda_1, \lambda_2, \dots, \lambda_n$,

eigenvalues of $A(v)$ are $\mu_1, \mu_2, \dots, \mu_{n-1}$, and

graph of A is G .

The λ - μ -SIEP: perturbing a diagonal entry

Theorem (K. H.M. and B.L. Shader 2014)

For given

$\lambda_1 < \mu_1 < \lambda_2 < \mu_2 < \dots < \mu_{n-1} < \lambda_n < \mu_n$: real numbers,

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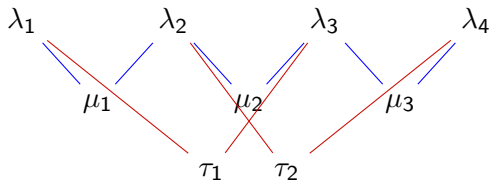
eigenvalues of $A + aE_{vv}$ are $\mu_1, \mu_2, \dots, \mu_n$, and

graph of A is G ,

where $a = \sum_i (\mu_i - \lambda_i)$.

A τ -pairing

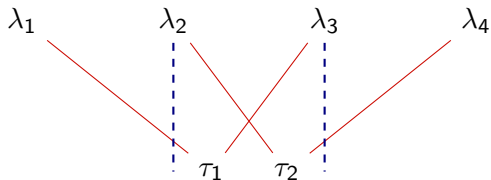
Recall the second order Cauchy interlacing inequalities:



If two consecutive τ 's are between two consecutive λ 's, it is called a τ -pairing

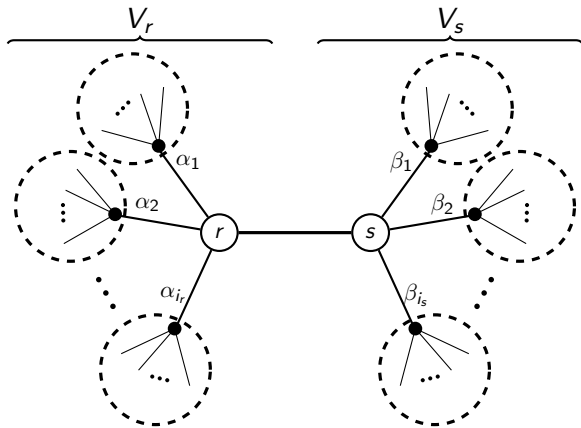
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Recall the second order Cauchy interlacing inequalities:



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A tree with adjacent vertices r and s



The λ - τ -SIEP: for trees

Theorem (K. H.M. and B.L. Shader 2014)

For given

$\lambda_1 < \lambda_2 < \dots < \lambda_n$ and $\tau_1 < \tau_2 < \dots < \tau_{n-2}$: real numbers,

G : a *tree* on n vertices, and

r, s : two vertices of G ,

where

$\lambda_i < \tau_i < \lambda_{i+2}$, and $\tau_i \neq \lambda_{i+1}$ for $i = 1, \dots, n - 2$,

there are k τ -pairings, and

$T[V_r \setminus \{r\}]$ and $T[V_s \setminus \{s\}]$ each have at least k vertices,

there is an $n \times n$ real symmetric matrix A such that

eigenvalues of A are $\lambda_1, \lambda_2, \dots, \lambda_n$,

eigenvalues of $A(\{r, s\})$ are $\tau_1, \tau_2, \dots, \tau_{n-2}$, and

graph of A is G .

The λ - τ -SIEP: for connected graphs

Theorem (K. H.M. and B.L. Shader 2014)

Under the same assumptions, if there are k τ -pairings, and G has a spanning tree T as before, then there is an $n \times n$ real symmetric matrix A such that

eigenvalues of A are $\lambda_1, \lambda_2, \dots, \lambda_n$,

eigenvalues of $A(\{r, s\})$ are $\tau_1, \tau_2, \dots, \tau_{n-2}$, and

graph of A is G .

The λ - τ -SIEP: perturbing two diagonal entries

Theorem (K. H.M. and B.L. Shader 2014)

Under the same assumptions, if there are k τ -pairings, and G has a spanning tree T as before, for given

$\lambda_1 < \lambda_2 < \cdots < \lambda_n$ and $\tau_1 < \tau_2 < \cdots < \tau_n$: real numbers,

where

$\lambda_i < \tau_i < \lambda_{i+2}$, and $\tau_i \neq \lambda_{i+1}$ for $i = 1, \dots, n-2$,

$\lambda_j < \tau_j$ for $j = n-1, n-2$,

there is an $n \times n$ real symmetric matrix A and real numbers a_r and a_s such that

eigenvalues of A are $\lambda_1, \lambda_2, \dots, \lambda_n$,

eigenvalues of $A + a_r E_{rr} + a_s E_{ss}$ are $\tau_1, \tau_2, \dots, \tau_n$, and

graph of A is G .

The Nowhere-zero Eigenbasis SIEP: for trees

Theorem (K. H.M. and B.L. Shader 2014)

For given

$\lambda_1 < \lambda_2 < \cdots < \lambda_n$: real numbers, and

G : a *tree* on n vertices,

there is an $n \times n$ real symmetric matrix A such that

eigenvalues of A are $\lambda_1 < \lambda_2 < \cdots < \lambda_n$,

graph of A is G , and

none of the eigenvectors of A has a zero entry.

The Nowhere-zero Eigenbasis SIEP: for connected graphs

Theorem (K. H.M. and B.L. Shader 2014)

For given

$\lambda_1 < \lambda_2 < \cdots < \lambda_n$: real numbers, and

G : a *connected graph* on n vertices,

there is an $n \times n$ real symmetric matrix A such that

eigenvalues of A are $\lambda_1 < \lambda_2 < \cdots < \lambda_n$,

graph of A is G , and

none of the eigenvectors of A has a zero entry.

Overcoming Difficulties

Derivatives of F

Consider the maps

$$F : M \mapsto (\lambda_1(M), \lambda_2(M), \dots, \lambda_n(M)),$$

where $\lambda_i(M)$ is the i -th smallest eigenvalue of M .

$$G : M \mapsto (c_0(M), c_1(M), \dots, c_{n-1}(M)),$$

where $c_i(M)$ is the coefficient of x^i in the characteristic polynomial of M .

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$$f : M \mapsto (\operatorname{tr} M, \operatorname{tr} M^2, \dots, \operatorname{tr} M^n).$$

Then,

$$\frac{\partial}{\partial x_t} (\operatorname{tr} M^k) = 2k \left(M^{k-1} \right)_{i,j}.$$

Then

$$\text{Jac}(f)|_A = \left[\begin{array}{ccc|ccc} I_{11} & \cdots & I_{nn} & I_{i_1 j_1} & \cdots & I_{i_m j_m} \\ A_{11} & \cdots & A_{nn} & A_{i_1 j_1} & \cdots & A_{i_m j_m} \\ \vdots & \ddots & \vdots & \vdots & \ddots & \vdots \\ A_{11}^{n-1} & \cdots & A_{nn}^{n-1} & A_{i_1 j_1}^{n-1} & \cdots & A_{i_m j_m}^{n-1} \end{array} \right],$$

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and

$$\text{Jac}_x(f)|_A = \left[\begin{array}{ccc|ccc} l_{11} & \cdots & l_{nn} & l_{i_1 j_1} & \cdots & l_{i_m j_m} \\ A_{11} & \cdots & A_{nn} & A_{i_1 j_1} & \cdots & A_{i_m j_m} \\ \vdots & \ddots & \vdots & \vdots & \ddots & \vdots \\ A_{11}^{n-1} & \cdots & A_{nn}^{n-1} & A_{i_1 j_1}^{n-1} & \cdots & A_{i_m j_m}^{n-1} \end{array} \right].$$

Nonsingularity of $\text{Jac}_x(\mathbf{f})|_A$

In the λ - μ -SIEP for connected graphs let A be a solution for the λ - μ -SIEP for a spanning tree T of G , $B = A(v)$, and let

$$M = \begin{bmatrix} 2x_1 & x_{n+1} & y_1 & \cdots & \\ x_{n+1} & 2x_2 & x_{n+2} & \ddots & \vdots \\ y_1 & x_{n+2} & 2x_3 & \ddots & y_m \\ \vdots & \ddots & \ddots & \ddots & x_{2n-1} \\ \cdots & y_m & x_{2n-1} & 2x_n & \end{bmatrix}, N = M(v).$$

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Define

$$f(\mathbf{x}, \mathbf{y}) := \left(\frac{\text{tr } M}{2}, \frac{\text{tr } M^2}{4}, \dots, \frac{\text{tr } M^n}{2n}, \frac{\text{tr } N}{2}, \frac{\text{tr } N^2}{4}, \dots, \frac{\text{tr } N^{n-1}}{2(n-1)} \right).$$

Let \tilde{B} be the matrix obtained from B by inserting a zero row and column in the ν -th row and column of it. Then

$$\text{Jac}_x(f) \Big|_A = \left[\begin{array}{ccc|ccc} I_{11} & \cdots & I_{nn} & I_{i_1 j_1} & \cdots & I_{i_{n-1} j_{n-1}} \\ A_{11} & \cdots & A_{nn} & A_{i_1 j_1} & \cdots & A_{i_{n-1} j_{n-1}} \\ \vdots & \ddots & \vdots & \vdots & \ddots & \vdots \\ A_{11}^{n-1} & \cdots & A_{nn}^{n-1} & A_{i_1 j_1}^{n-1} & \cdots & A_{i_{n-1} j_{n-1}}^{n-1} \\ \hline \tilde{I}_{11} & \cdots & \tilde{I}_{nn} & \tilde{I}_{i_1 j_1} & \cdots & \tilde{I}_{i_{n-1} j_{n-1}} \\ \tilde{B}_{11} & \cdots & \tilde{B}_{nn} & \tilde{B}_{i_1 j_1} & \cdots & \tilde{B}_{i_{n-1} j_{n-1}} \\ \vdots & \ddots & \vdots & \vdots & \ddots & \vdots \\ \tilde{B}_{11}^{n-2} & \cdots & \tilde{B}_{nn}^{n-2} & \tilde{B}_{i_1 j_1}^{n-2} & \cdots & \tilde{B}_{i_{n-1} j_{n-1}}^{n-2} \end{array} \right].$$

Let \tilde{B} be the matrix obtained from B by inserting a zero row and column in the v -th row and column of it. Then

$$\text{Jac}_x(f)\Big|_A = \left[\begin{array}{ccc|ccc} I_{11} & \cdots & I_{nn} & I_{i_1 j_1} & \cdots & I_{i_{n-1} j_{n-1}} \\ A_{11} & \cdots & A_{nn} & A_{i_1 j_1} & \cdots & A_{i_{n-1} j_{n-1}} \\ \vdots & \ddots & \vdots & \vdots & \ddots & \vdots \\ A_{11}^{n-1} & \cdots & A_{nn}^{n-1} & A_{i_1 j_1}^{n-1} & \cdots & A_{i_{n-1} j_{n-1}}^{n-1} \\ \hline \tilde{I}_{11} & \cdots & \tilde{I}_{nn} & \tilde{I}_{i_1 j_1} & \cdots & \tilde{I}_{i_{n-1} j_{n-1}} \\ \tilde{B}_{11} & \cdots & \tilde{B}_{nn} & \tilde{B}_{i_1 j_1} & \cdots & \tilde{B}_{i_{n-1} j_{n-1}} \\ \vdots & \ddots & \vdots & \vdots & \ddots & \vdots \\ \tilde{B}_{11}^{n-2} & \cdots & \tilde{B}_{nn}^{n-2} & \tilde{B}_{i_1 j_1}^{n-2} & \cdots & \tilde{B}_{i_{n-1} j_{n-1}}^{n-2} \end{array} \right].$$

$$\det(\text{Jac}_x(f)\Big|_A) = ?$$

Let \tilde{B} be the matrix obtained from B by inserting a zero row and column in the v -th row and column of it. Then

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$$\det(\text{Jac}_x(f)\Big|_A) = ?$$

Solution: Rows of $\text{Jac}_x(f)\Big|_A$ are linearly independent.

Assume $\alpha^T \text{Jac}_x(f)|_A = \mathbf{0}$. i.e.

$$\alpha_1 \text{Jac}_1 + \cdots + \alpha_{2n-1} \text{Jac}_{2n-1} = \mathbf{0} \quad (*)$$

Assume $\alpha^T \text{Jac}_x(f)|_A = \mathbf{0}$. i.e.

$$\alpha_1 \text{Jac}_1 + \cdots + \alpha_{2n-1} \text{Jac}_{2n-1} = \mathbf{0} \quad (*)$$

Define:

$$p(x) = \alpha_1 + \alpha_2 x + \cdots + \alpha_n x^{n-1},$$

$$q(x) = \alpha_{n+1} + \alpha_{n+2} x + \cdots + \alpha_{2n-1} x^{n-2}.$$

and

$$X = p(A) + \widetilde{q(B)}.$$

Assume $\alpha^T \text{Jac}_x(f)|_A = \mathbf{0}$. i.e.

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and

$$X = p(A) + \widetilde{q(B)}.$$

Then

$$(*) \iff \begin{cases} X \circ A = O, \\ X \circ I = O. \end{cases}$$

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Also note that $(AX - XA)(v) = O$.

Assume $\alpha^T \text{Jac}_x(f)|_A = \mathbf{0}$. i.e.

$$\alpha_1 \text{Jac}_1 + \cdots + \alpha_{2n-1} \text{Jac}_{2n-1} = \mathbf{0} \quad (*)$$

Define:

$$p(x) = \alpha_1 + \alpha_2 x + \cdots + \alpha_n x^{n-1},$$

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and

$$X = p(A) + \widetilde{q(B)}.$$

Then

$$(*) \iff \begin{cases} X \circ A = O, \\ X \circ I = O. \end{cases}$$

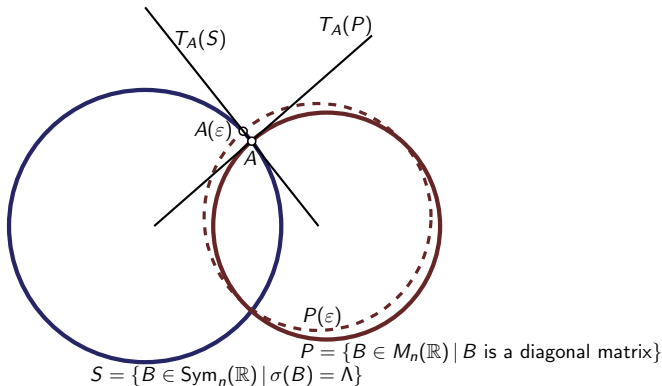
Also note that $(AX - XA)(v) = O$. It can be shown that $X = O$, and $p(x) \equiv 0$, $q(x) \equiv 0$.

Hence $\alpha = \mathbf{0}$.

Future
Work

Transverse Intersections

The mentioned problem can be described in terms of some manifolds intersecting transversally, which gives a more general approach to the Jacobian method.



Question: Could this help us to solve the cases where there are repeated eigenvalues, or the interlacing inequalities are not strict?

The Constant Rank Theorem vs. IFT

Theorem (Constant Rank Theorem¹)

Assume $a \in U \subseteq \mathbb{R}^n$, $F = (f_1, \dots, f_m) : U \rightarrow \mathbb{R}^m$ is $C^\infty(U)$, and the rank of $\text{Jac}(F)|_x$ is k for all x in a neighborhood of a . Then there are open neighborhoods V of a and W of $F(a)$ and diffeomorphisms $\phi : V \rightarrow \mathbb{R}^n$ and $\psi : W \rightarrow \mathbb{R}^m$ with

$$\begin{array}{ccc} V & \xrightarrow{F} & W \\ \phi \downarrow & & \downarrow \psi \\ \mathbb{R}^n & \longrightarrow & \mathbb{R}^m \end{array}$$

such that $\psi \circ F \circ \phi^{-1}(x_1, \dots, x_n) = (x_1, \dots, x_k, 0, \dots, 0)$.

Question: Could this help us to solve the cases where there are repeated eigenvalues, or the interlacing inequalities are not strict?

¹[S.G. Krantz and H.R. Parks, The Implicit Function Theorem: History, Theory, and Applications, 2013]

λ -SIEP for G when multiplicities are allowed

Question: What if the λ_i 's are not distinct?

$$A = \left[\begin{array}{cccccc} \times & \times & & & & \\ \times & \times & \times & & & \\ & \times & \times & \times & & \\ & & \times & \times & \times & \\ & & & \times & \times & \times \\ & & & & \times & \times \\ & & & & & \times \\ & & & & & \times \end{array} \right]$$

Λ

λ - μ -SIEP for G when multiplicities are allowed

Questions: What if some of these conditions are **not** necessary?

The λ_i 's are distinct.

The μ_i 's are distinct.

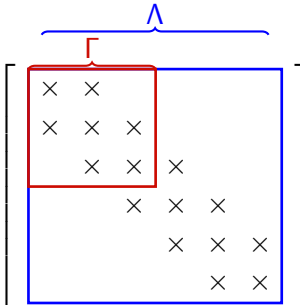
The μ_i 's **strictly** interlace the λ_i 's.

$$A = \left[\begin{array}{cccccc} \times & \times & & & & \\ \times & \times & \times & & & \\ & \times & \times & \times & & \\ & & \times & \times & \times & \\ & & & \times & \times & \times \\ & & & & \times & \times & \times \\ & & & & & \times & \times \end{array} \right]$$

The diagram shows a matrix A with 7 rows and 7 columns. A red box encloses the upper-left $M \times M$ submatrix, where $M=5$. A blue bracket above the matrix spans the first 5 columns and is labeled Λ . The matrix contains 'x' marks in the following positions (row, column): (1,1), (1,2), (2,1), (2,2), (2,3), (3,2), (3,3), (3,4), (4,3), (4,4), (4,5), (5,4), (5,5), (5,6), (5,7), (6,5), (6,6), (6,7), (7,6), (7,7).

Generalization: λ - γ -SIEP for G when multiplicities are allowed

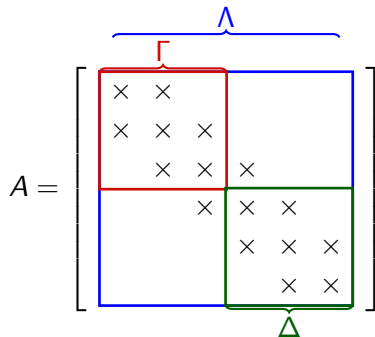
Question: What about the case that the eigenvalues of G and a $k \times k$ principal submatrix of it are prescribed?

$$A = \left[\begin{array}{cccc} \times & \times & & \\ \times & \times & \times & \\ & \times & \times & \times \\ & & \times & \times & \times \\ & & & \times & \times & \times \\ & & & & \times & \times \end{array} \right]$$


The diagram shows a matrix A with a red box around the top-left 3×3 submatrix, labeled Γ with a red bracket above it. A blue box around the top-left 6×6 submatrix is labeled Λ with a blue bracket above it. The matrix contains 'x' marks representing non-zero entries.

Generalization: λ - γ - δ -SIEP for G when multiplicities are allowed

Question: What about the case that the eigenvalues of G and a $k \times k$ principal submatrix of it and its complement are prescribed?



Other Problems

Question: Let G be a graph on n vertices and $\lambda_1 \leq \lambda_2 \leq \dots \leq \lambda_n$ be n real numbers. Is there a real symmetric matrix A such that $G(A) = G$ and $\lambda_k \in \sigma(A[1, 2, \dots, k])$, for $k = 1, 2, \dots, n$?

The diagram shows a 5x5 matrix A with entries marked by 'x'. The matrix is enclosed in a pink box labeled λ_5 . Inside it, a blue box labeled λ_4 covers the top-left 4x4 submatrix. A dark blue box labeled λ_3 covers the top-left 3x3 submatrix. A green box labeled λ_2 covers the top-left 2x2 submatrix. A red box labeled λ_1 covers the top-left 1x1 element. The matrix structure is as follows:

$$A = \begin{bmatrix} \times & \times & & & \\ \times & \times & \times & & \\ & \times & \times & \times & \\ & & \times & \times & \times \\ & & & \times & \times \end{bmatrix}$$



Thank **You!!**