

Normal and structured matrices under unitary structure-preserving transformations

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OUTLINE

- Introduction
- Structured matrices and structure-preserving transformations
- Jacobi-type algorithm for the reduction to the canonical form
- Convergence
- Finding the closest normal matrix with a given structure
- Numerical examples

E. Begović Kovač, H. Faßbender, P. Saltenberger: [On normal and structured matrices under unitary structure-preserving transformations](#). arXiv:1810.03369 [math.NA]

INTRODUCTION

- Set of normal matrices: $\mathcal{N} = \{X : XX^H = X^HX\}$
- X is normal if and only if there is unitary U such that

$$U^HXU = \begin{bmatrix} & & \\ & \diagdown & \\ & & \end{bmatrix}.$$

- *A. Ruhe: Closest normal matrix finally found!*
BIT 27 (4) (1987) 585–598.

Does NOT preserve given matrix structure.

INTRODUCTION

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Suppose that A has a structure \mathcal{S} , $A \in \mathcal{S}$.

Minimization problem:

$$\min \{ \|A - X\|_F^2 : X \in \mathcal{N} \cap \mathcal{S} \}$$

MAXIMIZATION PROBLEM

Theorem (Causey 1964, Gabriel 1979)

Let $A \in \mathbb{C}^{n \times n}$ and let $X = ZDZ^H$, where Z is unitary and D is diagonal. Then X is a nearest normal matrix to A in the Frobenius norm if and only if

$$(a) \quad \|\text{diag}(Z^H A Z)\|_F = \max_{Q Q^H = I} \|\text{diag}(Q^H A Q)\|_F, \text{ and}$$

$$(b) \quad D = \text{diag}(Z^H A Z).$$

→ Finding the closest normal matrix is equivalent to **finding an unitary transformation that maximizes Frobenius norm of the diagonal**.

→ This theorem has to be modified to fulfill structure-preserving requirements.

- N. J. Higham: *Matrix nearness problem and applications*.
In Applications of Matrix theory 22 (1989) 1–27.

Structured matrices and structure preserving transformations

STRUCTURED MATRICES

We study the following structures:

- **Hamiltonian** (J -Hermitian)
- **Skew-Hamiltonian** (J -skew-Hermitian)
- **Per-Hermitian** (R -Hermitian)
- **Perskew-Hermitian** (R -skew-Hermitian)

$$\text{where } J = \begin{bmatrix} 0 & I_n \\ -I_n & 0 \end{bmatrix} \quad \text{and} \quad R = \begin{bmatrix} 0 & \cdots & 0 & 1 \\ \vdots & & \ddots & 0 \\ 0 & \ddots & & \vdots \\ 1 & 0 & \cdots & 0 \end{bmatrix}.$$

STRUCTURED MATRICES

We study the following structures:

- **Hamiltonian** $\rightarrow (JA)^H = JA$
- **Skew-Hamiltonian** $\rightarrow (JA)^H = -JA$
- **Per-Hermitian** $\rightarrow (RA)^H = RA$
- **Perskew-Hermitian** $\rightarrow (RA)^H = -RA$

where $J = \begin{bmatrix} 0 & I_n \\ -I_n & 0 \end{bmatrix}$ and $R = \begin{bmatrix} 0 & \cdots & 0 & 1 \\ \vdots & & \ddots & 0 \\ 0 & \ddots & & \vdots \\ 1 & 0 & \cdots & 0 \end{bmatrix}$.

STRUCTURE-PRESERVING TRANSFORMATIONS

- For Hamiltonian and skew-Hamiltonian

→ J -unitary

- For per-Hermitian and perskew-Hermitian

→ R -unitary

STRUCTURE-PRESERVING TRANSFORMATIONS

- For Hamiltonian and skew-Hamiltonian

M is **symplectic** if $M^H J M = J$.

- For per-Hermitian and perskew-Hermitian

M is **perplectic** if $M^H R M = R$.

STRUCTURE-PRESERVING TRANSFORMATIONS

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M is **symplectic** if $M^H J M = J$.

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M is **perplectic** if $M^H R M = R$.

| manifold | tangent subspace at I | orthogonal subspace at I |
|--------------------------|----------------------------------|-----------------------------------|
| symplectic perplectic | Hamiltonian perskew-Hermitian | skew-Hamiltonian per-Hermitian |
| Lie group | Lie algebra | Jordan algebra |

Table: Geometric and algebraic setting for the structured matrices

HAMILTONIAN AND SKEW-HAMILTONIAN

- **Hamiltonian** A (J -Hermitian):

$$(JA)^H = JA, \quad \text{that is } A^H = JAJ, \quad \text{where } J = \begin{bmatrix} 0 & I \\ -I & 0 \end{bmatrix}.$$

We can write it as

$$A = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & -A_{11}^H \end{bmatrix}, \quad A_{12}^H = A_{12}, \quad A_{21}^H = A_{21}.$$

HAMILTONIAN AND SKEW-HAMILTONIAN

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- **Skew-Hamiltonian** A (J -skew-Hermitian):

$$(JA)^H = -JA, \quad \text{that is } A^H = -JAJ.$$

We can write it as

$$A = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{11}^H \end{bmatrix}, \quad A_{12}^H = -A_{12}, \quad A_{21}^H = -A_{21}.$$

- For every skew-Hamiltonian W there is Hamiltonian H (and viceversa) such that $W = \imath H$.

CANONICAL FORM — HAMILTONIAN

Theorem (BK, Faßbender, Saltenberger)

For every normal Hamiltonian $A \in \mathbb{C}^{2n \times 2n}$ there is unitary symplectic U such that

$$U^H A U = \begin{bmatrix} D_1 & 0 & 0 & 0 \\ 0 & D_2 & 0 & D_3 \\ 0 & 0 & -D_1^H & 0 \\ 0 & -D_3 & 0 & D_2 \end{bmatrix},$$

where D_j , $j = 1, 2, 3$ diagonal matrices,

$D_1 \in \mathbb{C}^{n_1 \times n_1}$, $D_2 \in \mathbb{R}^{n_2 \times n_2}$, $D_3 \in \mathbb{R}^{n_2 \times n_2}$, $n_1 + n_2 = n$.

$$U^H A U = \begin{bmatrix} \Lambda_1 & \Lambda_2 \\ -\Lambda_2 & -\Lambda_1^H \end{bmatrix} = \begin{bmatrix} \diagdown & \diagdown \\ \diagup & \diagup \end{bmatrix} =: \Lambda_{\mathcal{H}}$$

PER-HERMITIAN AND PERSKEW-HERMITIAN

- **Per-Hermitian** A (F -Hermitian):

$$(FA)^H = FA, \quad \text{that is } A^H = FAF,$$

where $F = \begin{bmatrix} 0 & \cdots & 0 & 1 \\ \vdots & & \ddots & 0 \\ 0 & \ddots & & \vdots \\ 1 & 0 & \cdots & 0 \end{bmatrix}.$

→ Hermitian about its anti-diagonal

- **Perskew-Hermitian** A (F -skew-Hermitian):

$$(FA)^H = -FA, \quad \text{that is } A^H = -FAF.$$

→ Skew-Hermitian about its anti-diagonal

- For every perskew-Hermitian K there is per-Hermitian M (and viceversa) such that $K = \imath M$.

CANONICAL FORM — PER-HERMITIAN

Theorem (BK, Faßbender, Saltenberger)

For every normal per-Hermitian $A \in \mathbb{C}^{2n \times 2n}$ there is unitary perplectic U such that

$$U^H A U = \begin{bmatrix} D_1 & 0 & 0 & 0 \\ 0 & D_2 & D_3 & 0 \\ 0 & F D_3 F & F D_2 F & 0 \\ 0 & 0 & 0 & F D_1 F \end{bmatrix},$$

where D_1 i D_2 are diagonal, and D_3 is antidiagonal matrix, $D_1 \in \mathbb{C}^{n_1 \times n_1}$, $D_2 \in \mathbb{R}^{n_2 \times n_2}$, $D_3 \in \mathbb{R}^{n_2 \times n_2}$, $n_1 + n_2 = n$.

$$U^H A U = \begin{bmatrix} \Lambda_1 & \Lambda_2 F \\ F \Lambda_2 & F \Lambda_1^H F \end{bmatrix} = \begin{bmatrix} \diagdown & \diagup \\ \diagup & \diagdown \end{bmatrix} =: \Lambda_{\mathcal{P}}$$

Jacobi-type algorithm
for the reduction to the canonical form

MAXIMIZATION ALGORITHM

$$\max_{ZZ^H=I, Z \in Sp_{2n}(\mathbb{C})} \{f_{\mathcal{H}}(Z) := \|\text{diag}(Z^H A Z)\|_F^2 + \|\text{diag}(J Z^H A Z)\|_F^2\}$$

- Iterative algorithm of the form

$$A^{(k+1)} = R_k^H A^{(k)} R_k, \quad k \geq 0.$$

- Transformations R_k are structure-preserving rotations obtained by embedding **two Jacobi rotations**

$$\begin{bmatrix} c & -s \\ s & c \end{bmatrix} := \begin{bmatrix} \cos \phi & -e^{i\alpha} \sin \phi \\ e^{-i\alpha} \sin \phi & \cos \phi \end{bmatrix} \quad \text{in } I_{2n}.$$

They are chosen to maximize

$$\|\text{diag}(A^{(k+1)})\|_F^2 + \|\text{diag}(J A^{(k+1)})\|_F^2.$$

- D. S. Mackey, N. Mackey, F. Tisseur: *Structured tools for structured matrices*. Electron. J. Linear Al. 10 (2003) 106–145.

MAXIMIZATION ALGORITHM

$$\max_{ZZ^H=I, Z \in P_{2n}(\mathbb{C})} \{f_{\mathcal{P}}(Z) := \|\text{diag}(Z^H A Z)\|_F^2 + \|\text{diag}(F Z^H A Z)\|_F^2\}$$

- Iterative algorithm of the form

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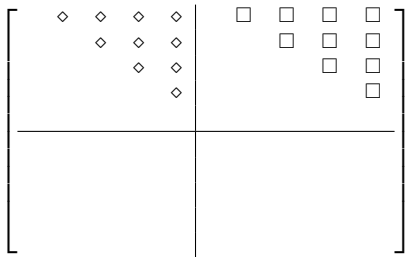
- D. S. Mackey, N. Mackey, F. Tisseur: *Structured tools for structured matrices*. Electron. J. Linear Al. 10 (2003) 106–145.

SYMPLECTIC ROTATIONS

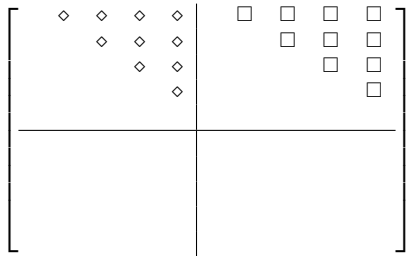
$$R(i, j, \phi, \alpha) = \left[\begin{array}{cc|cc} c & -s & & \\ \bar{s} & c & & \\ \hline & & c & -s \\ & & \bar{s} & c \end{array} \right] \begin{array}{l} i \\ j \\ n+i \\ n+j \end{array}$$

$$R(i, j, \phi, \alpha) = \left[\begin{array}{cc|cc} c & & & -s \\ & c & -\bar{s} & \\ \hline & s & c & \\ \bar{s} & & & c \end{array} \right] \begin{array}{l} i \\ j-n \\ n+i \\ j \end{array}$$

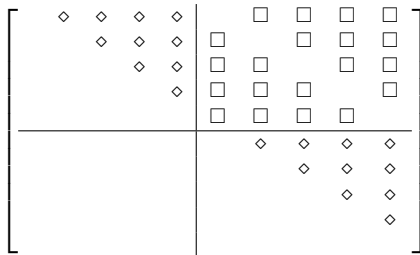
PIVOT POSITIONS (SYMPLECTIC)



PIVOT POSITIONS (SYMPLECTIC)



considering double rotations

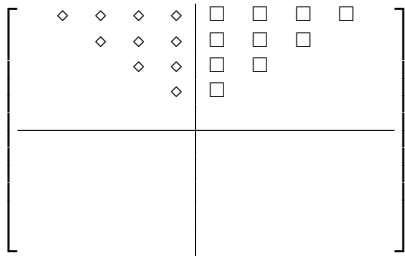


PERPLECTIC ROTATIONS

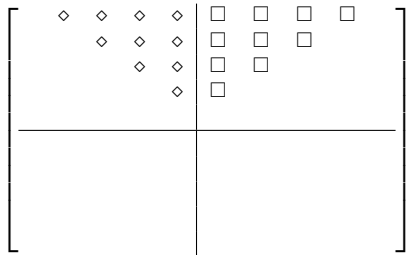
$$R(i, j, \phi, \alpha) = \left[\begin{array}{cc|cc} c & -s & & \\ \bar{s} & c & & \\ \hline & & c & \bar{s} \\ & & -s & c \end{array} \right] \begin{array}{l} i \\ j \\ 2n - j + 1 \\ 2n - i + 1 \end{array}$$

$$R(i, j, \phi, \alpha) = \left[\begin{array}{cc|cc} c & & -s & \\ & c & & \bar{s} \\ \hline \bar{s} & & c & \\ & -s & & c \end{array} \right] \begin{array}{l} i \\ 2n - j + 1 \\ j \\ 2n - i + 1 \end{array}$$

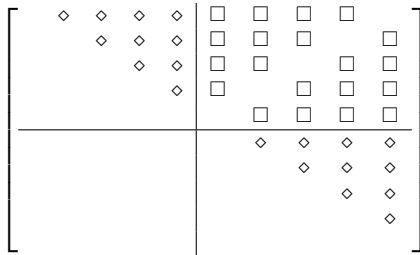
PIVOT POSITIONS (PERPLECTIC)



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considering double rotations



PIVOT PAIRS

- Pivot position $(i, j) \rightarrow$ cyclic pivot strategy
- Convergence condition:

$$|\langle \text{grad}f(Z), Z\dot{R}(i_k, j_k, 0, \alpha_k) \rangle| \geq \eta \|\text{grad}f(Z)\|_F,$$

where $\dot{R}(i, j, \phi, \alpha) = \frac{\partial}{\partial \phi} R(i, j, \phi, \alpha)$ and $f = f_{\mathcal{H}}$ or $f = f_{\mathcal{P}}$.

ROTATION ANGLES

- In step k we take ϕ_k and α_k such that $R_k = R(i_k, j_k, \phi_k, \alpha_k)$ maximizes

$$\|\text{diag}(A^{(k+1)})\|_F + \|P\text{diag}(PA^{(k+1)})\|_F,$$

for $P = J$ or $P = F$.

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for $P = J$ or $P = F$.

- Denote $A^{(k+1)} = A'$, $A^{(k)} = A$, $\phi_k = \phi$, $\alpha_k = \alpha$.
- For example, if A is Hamiltonian and we have symplectic rotation of the first type, we consider submatrix

$$A_{ij} = \begin{bmatrix} a_{ij} & a_{ij} & a_{i,n+i} & a_{i,n+j} \\ a_{ji} & a_{jj} & a_{j,n+i} & a_{j,n+j} \\ a_{n+i,i} & a_{n+i,j} & a_{n+i,n+i} & a_{n+i,n+j} \\ a_{n+j,i} & a_{n+j,j} & a_{n+j,n+i} & a_{n+j,n+j} \end{bmatrix}.$$

We have

$$A'_{ij} = \begin{bmatrix} \cos \phi & -e^{2\alpha} \sin \phi & 0 & 0 \\ e^{-2\alpha} \sin \phi & \cos \phi & 0 & 0 \\ 0 & 0 & \cos \phi & -e^{2\alpha} \sin \phi \\ 0 & 0 & e^{-2\alpha} \sin \phi & \cos \phi \end{bmatrix}^H A_{ij} \begin{bmatrix} \cos \phi & -e^{2\alpha} \sin \phi & 0 & 0 \\ e^{-2\alpha} \sin \phi & \cos \phi & 0 & 0 \\ 0 & 0 & \cos \phi & -e^{2\alpha} \sin \phi \\ 0 & 0 & e^{-2\alpha} \sin \phi & \cos \phi \end{bmatrix}.$$

ROTATION ANGLES—cont.

- We need

$$\begin{aligned} & |a'_{ii}|^2 + |a'_{jj}|^2 + |a'_{n+i,n+i}|^2 + |a'_{n+j,n+j}|^2 + \\ & + |a'_{i,n+i}|^2 + |a'_{j,n+j}|^2 + |a'_{n+i,i}|^2 + |a'_{n+j,j}|^2 \rightarrow \max. \end{aligned}$$

ROTATION ANGLES—cont.

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- Set $a_{rs} = x_{rs} + y_{rs}i$ and use the properties of a Hamiltonian matrix. We define function

$$\begin{aligned} g_{\mathcal{H}}(\phi, \alpha) = & 2|x'_{ii}|^2 + 2|y'_{ii}|^2 + 2|x'_{jj}|^2 + 2|y'_{jj}|^2 + \\ & + |x'_{i,n+i}|^2 + |y'_{i,n+i}|^2 + |x'_{j,n+j}|^2 + |y'_{j,n+j}|^2 \\ & + |x'_{n+i,i}|^2 + |y'_{n+i,i}|^2 + |x'_{n+j,j}|^2 + |y'_{n+j,j}|^2. \end{aligned}$$

- We take rotation angles ϕ and α that maximize $g_{\mathcal{H}}(\phi, \alpha)$.

REDUCTION TO CANONICAL FORM

Jacobi-type algorithm 1

Input: $A \in \mathbb{C}^{2n \times 2n} \in \mathcal{S}$, $Z_0 = I$

Output: structure-preserving unitary Z

REPEAT

 Select (i_k, j_k) .

 Find ϕ_k and α_k for $R(i_k, j_k, \phi_k, \alpha_k)$.

$$A^{(k+1)} = R_k^H A^{(k)} R_k$$

$$Z_{k+1} = Z_k R_k$$

UNTIL convergence

Convergence

CONVERGENCE

Theorem (BK, Faßbender, Saltenberger)

Let A be **Hamiltonian** (or skew-Hamiltonian) and let $(Z_k)_k$ be a sequence of **unitary symplectic** matrices generated by the Jacobi algorithm. Every accumulation point of $(Z_k)_k$ is a stationary point of function $f_{\mathcal{H}}$.

Theorem (BK, Faßbender, Saltenberger)

Let A be **per-Hermitian** (or perskew-Hermitian) and let $(Z_k)_k$ be a sequence of **unitary perplectic** matrices generated by the Jacobi algorithm. Every accumulation point of $(Z_k)_k$ is a stationary point of function $f_{\mathcal{P}}$.

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- M. Ishteva, P.-A. Absil, P. Van Dooren: *Jacobi algorithm for the best low multilinear rank approximation of symmetric tensors*. SIAM J. Matrix Anal. Appl. 34(2) (2013) 651–672.
- E. Begović Kovač, D. Kressner: *Structure-preserving low multilinear rank approximation of antisymmetric tensors*. SIAM. J. Matrix Anal. Appl. 38(3) (2017) 967–983.

THREE LEMMAS FOR $f_{\mathcal{H}}$

Lemma (BK, Faßbender, Saltenberger)

We have $\text{grad}f_{\mathcal{H}}(Z) = ZX$, where $\text{diag}(X) = 0$, $\text{diag}(JX) = 0$, and X is skew-Hermitian Hamiltonian.

Lemma (BK, Faßbender, Saltenberger)

For every unitary symplectic $Z \in \mathbb{C}^{2n \times 2n}$ there is symplectic rotation $R(i, j, \phi, \alpha)$ such that

$$|\langle \text{grad}f_{\mathcal{H}}(Z), Z\dot{R}(i, j, 0, \alpha) \rangle| \geq \eta \|\text{grad}f_{\mathcal{H}}(Z)\|_F, \quad \eta = \frac{4}{\sqrt{4n^2 - 4n}}.$$

Lemma (BK, Faßbender, Saltenberger)

Let $\hat{Z} \in \mathbb{C}^{2n \times 2n}$ be symplectic. Let $f = f_{\mathcal{H}}$ and $(Z_k, k \geq 0)$ be a sequence of unitary symplectic matrices generated by the Jacobi algorithm. If $\text{grad}f(\hat{Z}) \neq 0$, there exist $\epsilon > 0$ and $\delta > 0$ such that

$$\|Z_k - \hat{Z}\|_F < \epsilon \quad \Rightarrow \quad f(Z_{k+1}) - f(Z_k) \geq \delta.$$

Finding the closest normal matrix
with a given structure

THE CLOSEST NORMAL MATRIX

- Let A be Hamiltonian. Analogy with unstructured case:

- (i) Find Z that maximizes

$$f_{\mathcal{H}}(Z) = \|\text{diag}(Z^H A Z)\|_F^2 + \|\text{diag}(JZ^H A Z)\|_F^2,$$

- (ii) Extract the canonical form,

- (iii) Solution is given by $X = Z \begin{bmatrix} \diagdown & \diagdown \\ \diagup & \diagup \end{bmatrix} Z^H$.

→ But this can produce a matrix that is not normal!

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→ But this can produce a matrix that is not normal!

- We set

$$f_{\mathcal{D}}(Z) = \|\text{diag}(Z^H A Z)\|_F^2.$$

- (i) Find Z that maximizes $f_{\mathcal{D}}$.

- (ii) Extract the diagonal.

- (iii) Solution is given by $X = Z \begin{bmatrix} \diagdown & & \\ & & \\ & & \diagdown \end{bmatrix} Z^H.$

ADDITIONAL ROTATIONS

→ To find Z that maximizes $f_{\mathcal{D}}$ we add new rotations to the Jacobi algorithm.

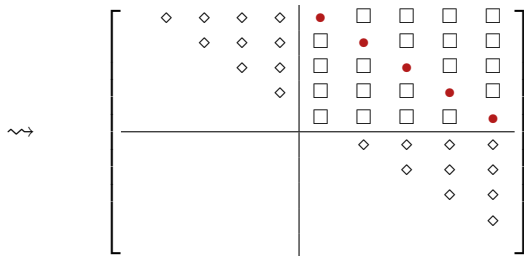
- Symplectic rotations

$$R(i, n+i, \phi, 0) = \begin{bmatrix} \cos \phi & -\sin \phi & & \\ & \sin \phi & & \\ & & \cos \phi & \\ & & & \sin \phi \end{bmatrix} \begin{matrix} i \\ \\ n+i \\ \end{matrix}$$

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- Perplectic rotations

$$R(i, 2n-i+1, \phi, -\frac{\pi}{2}) = \begin{bmatrix} \cos \phi & \imath \sin \phi \\ \imath \sin \phi & \cos \phi \end{bmatrix} \begin{matrix} i \\ 2n-i+1 \end{matrix}$$

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$$\rightsquigarrow \left[\begin{array}{cccc|cccc} \diamond & \diamond & \diamond & \diamond & \square & \square & \square & \square & \bullet \\ & & \diamond & \diamond & \square & \square & \square & \bullet & \square \\ & & & \diamond & \square & \square & \bullet & \square & \square \\ & & & & \square & \bullet & \square & \square & \square \\ & & & & \bullet & \square & \square & \square & \square \\ \hline & & & & & \diamond & \diamond & \diamond & \diamond \\ & & & & & & \diamond & \diamond & \diamond \\ & & & & & & & \diamond & \diamond \\ & & & & & & & & \diamond \end{array} \right]$$

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DIAGONALIZATION ALGORITHM

Jacobi-type algorithm 2

Input: $A \in \mathbb{C}^{2n \times 2n} \in \mathcal{S}$, $Z_0 = I$

Output: structure-preserving unitary Z

REPEAT

 Select (i_k, j_k) . (additional pivot positions are included)

 Find ϕ_k and α_k for $R(i_k, j_k, \phi_k, \alpha_k)$.

$$A^{(k+1)} = R_k^H A^{(k)} R_k$$

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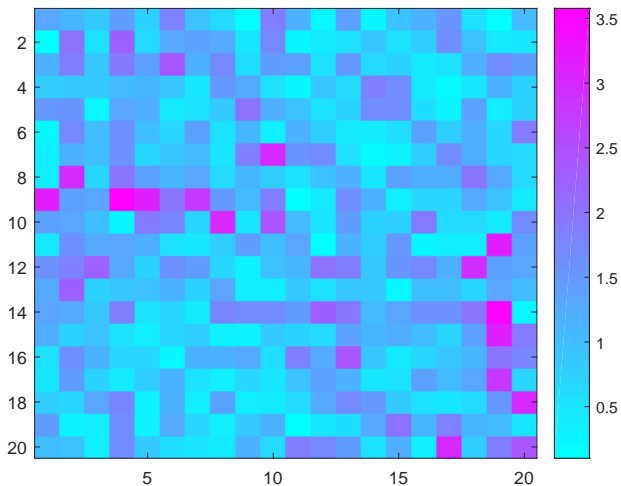
UNTIL convergence

Theorem (BK, Faßbender, Saltenberger)

Let A be **Hamiltonian** and let $(Z_k)_k$ be a sequence of **unitary symplectic** matrices generated by the Jacobi algorithm with additional rotations. Every accumulation point of $(Z_k)_k$ is a stationary point of function $f_{\mathcal{D}}$.

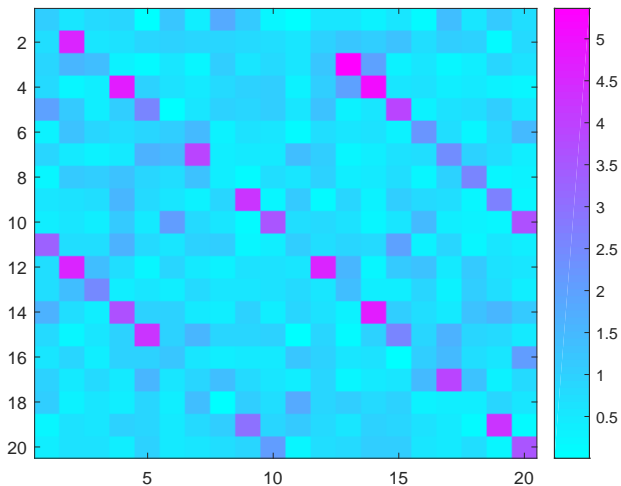
Numerical examples

NUMERICAL EXAMPLES — Canonical form



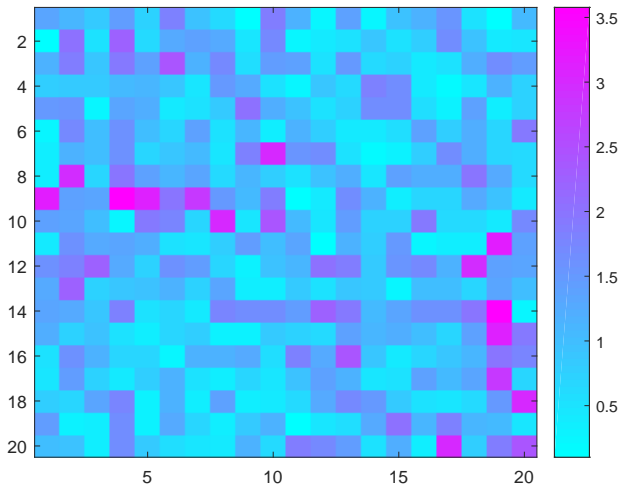
Random Hamiltonian 20×20 matrix.

NUMERICAL EXAMPLES — Canonical form



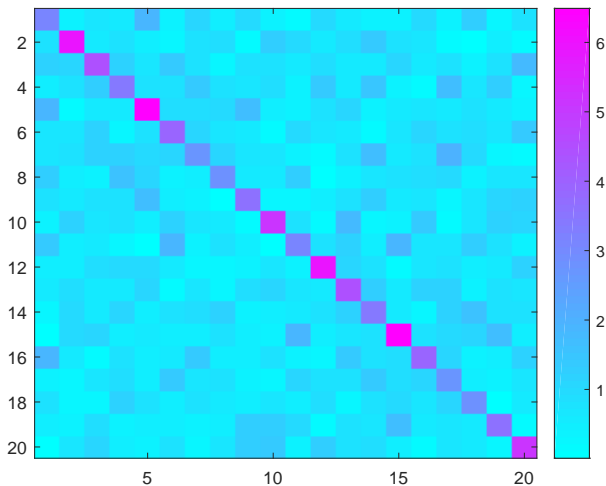
Reduction to the canonical form (Algorithm 1) after 10 cycles.

NUMERICAL EXAMPLES — Diagonalization



The same Hamiltonian 20×20 matrix.

NUMERICAL EXAMPLES — Diagonalization

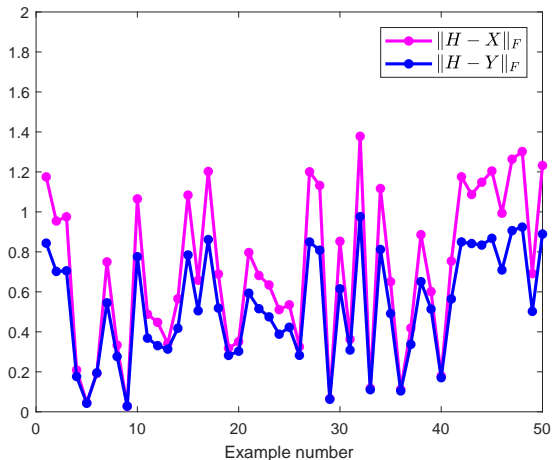


Diagonalization (Algorithm 2) after 10 cycles.

NUMERICAL EXAMPLES — Distance from normal matrix

We take normal Hamiltonian X and set $H = X + E$, such that H is Hamiltonian, but not normal.

Algorithm 2 on H gives its closest normal Y .



NUMERICAL EXAMPLES — Departure from normality

For any matrix A its Schur form

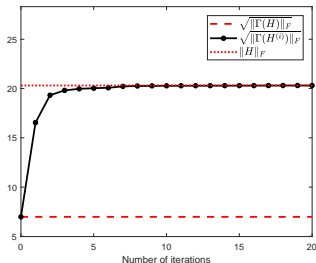
$$U^H A U = T = D + N$$

exists, where U is unitary, $D = \text{diag}(T)$ and N is strictly upper triangular. The quantity $\Delta(A) = \|N\|_F$ is referred to as A 's departure from normality.

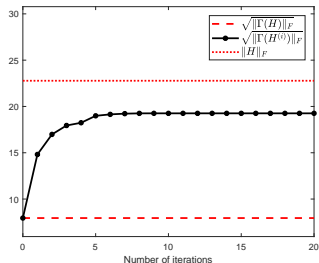
We compare $\Delta(H)$ and $\text{off}(H^{(20)})$ where $H^{(20)}$ is obtained by 20 iterations of Algorithm 2 and $\text{off}(A) = \|A - \text{diag}(A)\|_F^2$.

| Example i | Size of H_i | $\Delta(H_i)$ | $\text{off}(H_i^{(20)})$ |
|-------------|---------------|---------------------|--------------------------|
| 1 | 10 | $7.1 \cdot 10^{+0}$ | $6.4 \cdot 10^{+0}$ |
| 2 | 10 | $4.0 \cdot 10^{-3}$ | $3.1 \cdot 10^{-3}$ |
| 3 | 20 | $3.5 \cdot 10^{-5}$ | $3.1 \cdot 10^{-5}$ |
| 4 | 20 | $5.3 \cdot 10^{+2}$ | $4.4 \cdot 10^{+2}$ |
| 5 | 30 | $7.7 \cdot 10^{+0}$ | $6.7 \cdot 10^{+0}$ |
| 6 | 30 | $1.0 \cdot 10^{-1}$ | $9.0 \cdot 10^{-2}$ |
| 7 | 40 | $7.9 \cdot 10^{-7}$ | $6.6 \cdot 10^{-7}$ |
| 8 | 40 | $3.1 \cdot 10^{+3}$ | $2.7 \cdot 10^{+3}$ |
| 9 | 50 | $1.1 \cdot 10^{-2}$ | $9.5 \cdot 10^{-3}$ |
| 10 | 100 | $7.8 \cdot 10^{-7}$ | $6.8 \cdot 10^{-7}$ |

NUMERICAL EXAMPLES — Convergence of Algorithm 1



Normal Hamiltonian 20×20



Random Hamiltonian 20×20

$$\Gamma(A) := \|\text{diag}(Z^H A Z)\|_F^2 + \|\text{diag}(J Z^H A Z)\|_F^2$$

QUESTIONS???

