Asymptotic diagonalization of Hilbert-Schmidt operators

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Joint work with N. RAYMOND

Journée de l'Équipe d'Analyse Numérique 15 March 2018

A little history

Hint around the power method : Consider $A \in M_n(\mathbb{C}), x_0, y_0 \in \mathbb{C}^n$. For *z* sufficiently large : $=:s_{\nu}$

$$f(z) = \langle x_0, (zI - A)^{-1} y_0 \rangle = \sum_{\nu=0}^{+\infty} \underbrace{\overline{\langle x_0, A^{\nu} y_0 \rangle}}_{z^{\nu+1}} = \frac{\langle x_0, \mathsf{adj}(zI - A) y_0 \rangle}{\det(zI - A)}$$

Under some assumptions on x_0, y_0 and on A:

 $\frac{s_{\nu+1}}{s_{\nu}} \underset{\nu \to \infty}{\longrightarrow} \lambda_1, \text{ eigenvalue of maximum modulus.}$

More generaly : Being given the sequence of moments (s_{ν}) , find all the eigenvalues of a matrix A ?

- 1892 Hadamard extended the above idea to obtain any of the pole of a meromorphic function from its moments, using the sequence of Hankel determinants of the form $|s_{\nu+i+j}|_{0 \le i,j \le k}$.
- 1931 Aitken rediscovered a more algorithmic version of the method.
- 1954 Rutishauser introduced qd (quotient-difference) algorithm, avoiding Hankel determinants, reducing ill-posedness of the method, and then reformulated (first for tridiagonal matrices) into the LR algorithm (Gauss LU decomposition)





Rutishause

$$L_k R_k := A_k, \quad A_{k+1} := R_k L$$

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Hadamard



Rutishauser

1961 Francis, Kublanovskaya : QR algorithm

$$Q_k R_k := A_k, \quad A_{k+1} := R_k Q_k$$
$$A_{k+1} = Q_k^* A_k Q_k = (Q_1 \dots Q_k)^* A(Q_1 \dots Q_k)$$
spec A_k = spec A

Francis, Wilkinson : Fast variants with shifts

$$Q_k R_k := A_k - \sigma_k I_n, \quad A_{k+1} := R_k Q_k + \sigma_k I_n$$

1968 Colette Lebaud :

Remarques sur la convergence de la méthode Q.R., Publications mathématiques et informatique de Rennes, Tome (1967-1968), Exposé no. 5, p.1–17.

PhD Univ. Rennes : Contribution à l'étude de l'algorithme QR, 1971.



Kublanovskaya



Francis



Common convergence results for the QR algorithm

Theorem

Let $A \in M_n(\mathbb{C})$ and $P \in GL_n(\mathbb{C})$ such that $PAP^{-1} = diag(\lambda_1, \dots, \lambda_n) = D$ with $|\lambda_1| > |\lambda_2| > \dots > |\lambda_n| > 0.$

Assume moreover that P admits a LU factorization. Then

$$A_k \underset{k \to \infty}{\longrightarrow} \begin{pmatrix} \lambda_1 & (\star)_k \\ & \ddots \\ (0) & \lambda_n \end{pmatrix}.$$

The lower part converge as $\mathcal{O}(\mu^k)$ with $\mu = \max_{1 \le \ell \le n-1} \left| \frac{\lambda_{\ell+1}}{\lambda_\ell} \right|.$

- In presence of "same modulus" distinct eigenvalues: convergence "by block".
- Iterations preserve symmetry and Hessenberg structure.
- Convergence is quadratic for the Wilkinson shifted version, and then cubic for symmetric matrices.
- The algorithm is equivalent to the orthogonal simultaneous power iteration (this makes the assumption P = LU and improved convergence rates more intelligible)

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Continuous dynamical systems

- 1988 Chu and Norris, *Isospectral flows and abstract matrix factorizations*, SIAM J. Numer. Anal.
- 1991 Brockett, Dynamical systems that sort lists, diagonalize matrices and solve linear programming problems, Linear Algebra Appl.
- 1994 Wegner, Flow equations for Hamiltonians, Ann. Physik.
- 2010 Bach and Bru, *Rigorous foundations of the Brockett-Wegner flow for operators*, J. Evol. Equ.
- 2016 Bach and Bru, *Diagonalizing quadratic bosonic operators by non-autonomous flow equation*, Memoirs AMS.

References for some history :

Gutknecht and Parlett, From qd to LR, or, How were the qd and LR algorithms discovered?, IMA J. Numer. Anal. 31 (2011).

Golub and van der Vorst, *Eigenvalue computation in the 20th century*, J. Comput. Appl. Math. 123 (2000).

Our aim :

Get a unified proof for different bracket flow ODEs : H' = [H, G(H)], acting on Hilbert-Schmidt operators, and understand the QR method !

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Introduction

Main result

Bracket flows

Global existence and isospectrality Quadratic bracket flows Several examples

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A more specific result Relation with the discrete QR algorithm Numerical examples

Some notations

 $\begin{array}{l} \mathcal{L}(\mathcal{H}) \mbox{ bounded operators on a separable Hilbert space } \mathcal{H} \\ \mathcal{S}(\mathcal{H}), \, \mathcal{A}(\mathcal{H}), \, \mathcal{U}(\mathcal{H}) \mbox{ symmetric, skew-symmetric and unitary operators respectively} \end{array} }$

 $\mathcal{L}_2(\mathcal{H})$ Hilbert-Schmidt (compact) operators on \mathcal{H} , with the norm $\|\cdot\|_{HS}$:

$$||H||_{\mathrm{HS}}^{2} = \mathrm{tr}(H^{\star}H) = \sum_{n \ge 0} ||He_{n}||^{2} = \sum_{n=0}^{+\infty} |\lambda_{n}(H)|^{2}$$

 $\begin{aligned} \mathcal{S}_2(\mathcal{H}) \ &= \mathcal{S}(\mathcal{H}) \cap \mathcal{L}_2(\mathcal{H}) \\ \mathcal{A}_2(\mathcal{H}) \ &= \mathcal{A}(\mathcal{H}) \cap \mathcal{L}_2(\mathcal{H}) \end{aligned}$

 $(e_n)_{n\in\mathbb{N}}$ a Hilbert basis of $\mathcal H$

 $h_{i,j} = \langle He_i, e_j
angle$, for any $H \in \mathcal{L}(\mathcal{H})$

 $\mathcal{D}(\mathcal{H})$ bounded diagonal operators with respect to the Hilbert basis:

 $T \in \mathcal{D}(\mathcal{H}) \Leftrightarrow \forall n \in \mathbb{N}, \ \exists \lambda_n \in \mathbb{R}, \ Te_n = \lambda_n e_n$

 $\begin{array}{l} E_{i,j} \ (i \leq j) \text{ canonical Hilbert basis of } \mathcal{S}_2(\mathcal{H}) \\ E_{i,j}^{\pm} \ (i < j) \text{ canonical Hilbert basis of } \mathcal{A}_2(\mathcal{H}) \end{array}$

Main result

Consider $H \in \mathcal{C}^1(\mathbb{R}, \mathcal{L}_2(\mathcal{H}))$ such that

- i) the function $t \mapsto \|H(t)\|_{\text{HS}}$ is bounded,
- ii) the family $(h_{i,j})_{i,j\in\mathbb{N}}$ is balanced (symmetry-like property, uniform in *t*)
- iii) there exists a pointwise bounded and balanced family of measurable functions $(g_{i,j})_{i,j\in\mathbb{N}}$ defined on \mathbb{R} such that

$$\forall t \in \mathbb{R}, \ \forall i \in \mathbb{N}, \quad h'_{i,i}(t) = \sum_{j=0}^{+\infty} g_{i,j}(t) |h_{i,j}(t)|^2 \,.$$

Suppose that there exists T > 0 and a sign sequence $(\epsilon_\ell)_{\ell \in \mathbb{N}} \in \{-1, 1\}^{\mathbb{N}}$, such that

$$\forall t \geq T, \; \forall k, \ell \in \mathbb{N}, \; k \geq \ell, \; \epsilon_\ell g_{\ell,k}(t) \geq 0 \,. \tag{sign}$$

Then we have the integrability property

$$\forall \ell \in \mathbb{N}, \quad \sum_{j=0}^{+\infty} \int_T^{+\infty} |g_{\ell,j}(t)| |h_{\ell,j}(t)|^2 \mathrm{d}t < +\infty \,,$$

and the diagonal terms $h_{\ell,\ell}(t)$ converges at infinity.

Ideas of the proof

- Proof by induction on $\ell \in \mathbb{N}$, using monotonicity arguments
- Limit of the $h_{\ell,\ell}$ coefficient and integrability of the series is controled from the next lemma

Lemma

Consider $h \in C^1(\mathbb{R}_+)$ a real-valued bounded function, $F \in L^1(\mathbb{R}_+)$ and G a nonnegative measurable function defined on \mathbb{R}_+ such that

$$\forall t \ge 0, \ h'(t) = F(t) + G(t).$$

Then, the function G is integrable and h converges to a finite limit at infinity.

$$\int_{0}^{x} G(t) dt = h(x) - h(0) - \int_{0}^{x} F(t) dt \leq 2 \sup_{\mathbb{R}_{+}} |h| + \int_{0}^{+\infty} |F(t)| dt.$$

Therefore $G \in L^{1}(\mathbb{R}_{+})$ and $\lim_{x \to +\infty} h(x) = h(0) + \int_{0}^{+\infty} (F(t) + G(t)) dt.$

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Therefore $G\in L^1(\mathbb{R}_+)$ and $\lim_{x\to+\infty}h(x)=h(0)+\int_0^{+\infty}(F(t)+G(t))\mathrm{d}t$.

Main result (2)

Under the additional semi-uniform lower bound condition

$$\forall \ell \in \mathbb{N}, \ \exists c_{\ell} > 0, \ \forall j \neq \ell, \ \forall t \ge T, \ |g_{\ell,j}(t)| \ge c_{\ell},$$
 (lower bound)

we have the stronger integrability result

$$\sum_{j \neq \ell} \int_{T}^{+\infty} |h_{\ell,j}(t)|^2 dt = \int_{T}^{+\infty} \|H(t)e_{\ell} - h_{\ell,\ell}(t)e_{\ell}\|^2 \mathrm{d}t < +\infty \,.$$

Suppose moreover that, for any $\ell \in \mathbb{N}$, the function $t \mapsto ||H'(t)e_{\ell}||$ is bounded, then for any $\ell \in \mathbb{N}$, $H(t)e_{\ell}$ converges to $h_{\ell,\ell}(\infty)e_{\ell}$.

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Theorem

Suppose $G: S(H) \to A(H)$ is locally Lipschitz, then any solution to the following Cauchy problem

 $H' = [H, G(H)], \quad H(0) = H_0 \in S_2(\mathcal{H}).$

admits a unique global solution $H \in C^1(\mathbb{R}, S_2(\mathcal{H}))$ and there exists $U \in C^1(\mathbb{R}, \mathcal{U}(\mathcal{H}))$ such that

 $H(t) = U(t)^* H_0 U(t), \quad t \in \mathbb{R}.$

Proof: As long H(t) is defined, consider the following Cauchy problem:

U'(t) = U(t)G(H(t)), $U(0) = \mathrm{Id}.$

Then $U(t) \in \mathcal{U}(\mathcal{H})$ and $H(t) = U^{\star}(t)H_0U(t)$.

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Then $U(t) \in \mathcal{U}(\mathcal{H})$ and $H(t) = U^{*}(t)H_{0}U(t)$.

$$\frac{d}{dt}(UU^{\star}) = U'U^{\star} + UU'^{\star}$$
$$= UG(H)U^{\star} + U(U\underbrace{G(H)}_{\in \mathcal{A}(\mathcal{H})})^{\star} = UG(H)U^{\star} - UG(H)U^{\star} = 0.$$

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$$\begin{split} \frac{d}{dt}(UHU^{\star}) &= U'HU^{\star} + UH'U^{\star} + UHU'^{\star} \\ &= UG(H)HU^{\star} + U[H,G(H)]U^{\star} - UHG(H)U^{\star} \\ &= U\Big(G(H)H - HG(H) + [H,G(H)]\Big)U^{\star} = 0. \end{split}$$

Quadratic bracket flows

Consider H' = [H, G(H)], in the case $G \in \mathcal{L} (\mathcal{S}_2(\mathcal{H}), \mathcal{A}_2(\mathcal{H}))$.

The map G is said to be diagonalizable when there exists a Hilbertian basis $(e_n)_{n \in \mathbb{N}}$ and a skew-symmetric family $(g_{i,j})$ such that $G(E_{i,j}) = g_{i,j}E_{i,j}^{\pm}$.

We then have

$$h'_{i,i}(t) = -2\sum_{j} g_{i,j} |h_{i,j}(t)|^2,$$

and prove from the main result :

Corollary

Let $G \in \mathcal{L} (S_2(\mathcal{H}), \mathcal{A}_2(\mathcal{H}))$ be a diagonalizable map such that its matrix-eigenvalue satisfies (sign) and (lower bound). Then the unique global solution $H \in \mathcal{C}^1(\mathbb{R}, S_2(\mathcal{H}))$ converges weakly in $S_2(\mathcal{H})$ to some $H_\infty \in \mathcal{D}_2(\mathcal{H})$. Moreover, any diagonal term α of H_∞ with multiplicity m is an eigenvalue of H_0 of multiplicity at least m.

$$G_{\mathsf{Br}}(H) = [H, A], \quad \text{where } A = \mathsf{diag}(a_1, \ldots) \in \mathcal{D}(\mathcal{H}), \quad a_1 > a_2 > \ldots > 0$$

• skew-symmetric matrix-eigenvalue :

$$g_{i,j} = a_j - a_i$$

• (sign) assumption :

$$\forall k, \ell \in \mathbb{N}, \ k \ge \ell, \ -g_{\ell,k} \ge 0$$

• (lower bound) assumption :

$$\forall \ell \in \mathbb{N}, \ \forall j \neq \ell, \ |g_{\ell,j}| \ge \min(|a_\ell - a_{\ell-1}|, |a_\ell - a_{\ell+1}|)$$

Example 2: Toda's choice

$$G_{\text{Tod}}(H) = H^{-} - (H^{-})^{*}, \qquad H^{-} = \sum_{1 \le i \le j} h_{i,j} e_{i}^{*} e_{j}$$

• skew-symmetric matrix-eigenvalue :

$$g_{i,j} = -1, \quad i < j$$

• (sign) assumption :

$$\forall k, \ell \in \mathbb{N}, \ k \ge \ell, \ -g_{\ell,k} \ge 0$$

• (lower bound) assumption :

$$\forall \ell \in \mathbb{N}, \ \forall j \neq \ell, \ |g_{\ell,j}| \ge 1$$

 $G_{\mathsf{Weg}}(H) = [H, \mathsf{diag}(H)]$

- $G_{Weg} \notin \mathcal{L}(\mathcal{S}_2(\mathcal{H}), \mathcal{A}_2(\mathcal{H}))$
- but $G_{Weg}(S_2(\mathcal{H})) \subset \mathcal{A}_2(\mathcal{H})$ therefore the flow is unitarily equivalent and global.
- diagonal terms follow the dynamic:

$$h'_{i,i} = \sum_{j=0}^{+\infty} \underbrace{(h_{i,i} - h_{j,j})}_{g_{i,j}(t)} |h_{ij}|^2.$$

• What about (sign) and (lower bound) assumptions ? If for large time, the diagonal terms $(h_{\ell,\ell})$ become "sufficiently" distinct, both assumptions are valid.

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A more specific result Relation with the discrete QR algorithm Numerical examples More about the Toda flow : $G(H) = H^- - (H^-)^*$.

Proposition

Assume that $H_0 \in \mathcal{L}_2(\mathcal{H})$ (not necessarily symmetric) is diagonalizable with eigenvalues $(\lambda_j)_{j\geq 0}$ such that

 $\mathsf{Re}\lambda_0 > \mathsf{Re}\lambda_1 > \dots$

Let us also assume that we may find $P \in \mathcal{L}(\mathcal{H})$ such that $PH_0P^{-1} = \text{diag}(\lambda_j)$ and such that all the minors of $P : P_J = (\langle Pe_i, e_j \rangle)_{0 \le i,j \le J}$ are invertible.

For $\ell \in \mathbb{N}$, we denote $\delta_\ell = \min_{0 \leq j \leq \ell} \mathsf{Re}(\lambda_j - \lambda_{j+1})$, then

$$H(t)e_{\ell} - \sum_{j=\ell+1}^{+\infty} \langle H(t)e_{\ell}, e_j \rangle e_j = \lambda_{\ell}e_{\ell} + \mathcal{O}\left(e^{-t\delta_{\ell}}\right) \,.$$

- · Find exponentially fast eigenvalues of any Hilbert-Schmidt operator
- If H(0) is symmetric, then H(t) converges to a diagonal operator H_{∞} that has exactly the same eigenvalues as H_0
- Even in infinite dimension no eigenvalue is lost in the limit !

1988 Chu and Norris proposed an abstract decomposition of the flow

 $H'(t) = \left[H(t), G_{\mathsf{Tod}}(H(t))\right], \qquad H(0) = H_0 \in \mathcal{L}_2(\mathcal{H}),$

$$\begin{split} g_1'(t) &= g_1(t) G_{\mathsf{Tod}}(H(t)) \,, & g_1(0) = \mathrm{Id} \,, \\ g_2'(t) &= (H(t) - G_{\mathsf{Tod}}(H(t))) g_2(t) \,, & g_2(0) = \mathrm{Id} \,. \end{split}$$

so that

$$H(t) = g_1(t)^{-1} H_0 g_1(t) = g_2(t) H_0 g_2^{-1}(t)$$

$$e^{tH_0} = g_1(t)g_2(t), \qquad e^{tH(t)} = g_2(t)g_1(t).$$

- $G_{\mathsf{Tod}}(H) = H_{-} H_{-}^{\star} \in \mathcal{A}(\mathcal{H})$ therefore $g_1(t) \in \mathcal{U}(\mathcal{H})$
- $H G_{\mathsf{Tod}}(H) = H_+ + H_-^*$ is upper triangular and so is $g_2(t)$
- at t = 1: e^{H₀} = g₁(1)g₂(1) is a QR factorization of e^{H₀} and e^{H(1)} = g₂(1)g₁(1) is then the first iterate of the QR algorith
- Toda flow ≃ sampling of the discrete algorithm (up to a change in g₂(0) to ensure corresponding normalization choices in the diagonal of g₂(1))
- spec(e^{H0}) = {e^{λj}} with descreasing moduli (convergence rate)
- Remark : this can be done for many matrix factorization : QR, LU, Cholesky.

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- Toda flow \simeq sampling of the discrete algorithm (up to a change in $g_2(0)$ to ensure corresponding normalization choices in the diagonal of $g_2(1)$)
- spec(e²⁰) = {e^j} with descreasing moduli (convergence rate)
- Remark : this can be done for many matrix factorization : QR, LU, Cholesky.

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- $H G_{\mathsf{Tod}}(H) = H_+ + H_-^*$ is upper triangular and so is $g_2(t)$

• at t = 1: $e^{H_0} = g_1(1)g_2(1)$ is a QR factorization of e^{H_0} and $e^{H(1)} = g_2(1)g_1(1)$ is then the first iterate of the QR algorithm

- Toda flow \simeq sampling of the discrete algorithm (up to a change in $g_2(0)$ to ensure corresponding normalization choices in the diagonal of $g_2(1)$)
- spec $(e^{H_0}) = \{e^{\lambda_j}\}$ with descreasing moduli (convergence rate)
- Remark : this can be done for many matrix factorization : QR, LU, Cholesky.

Numerical examples

Finite dimension case :

- strong convergence to the limit
- alternative : compactness of $\mathcal{U}(\mathcal{H})$
- linearization of the flow in $\mathcal{U}(\mathcal{H})$ around $U_\infty,$ parameterized by its Lie algebra $\mathcal{A}(\mathcal{H})$
- \Rightarrow exponential convergence, whatever is the stable manifold in consideration.

Datas for the computations :

• Initial data :
$$H_0 = Q \begin{pmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 4 & 0 & 0 & 0 \\ 0 & 0 & 0 & 16 & 0 \\ 0 & 0 & 0 & 0 & 25 \end{pmatrix} Q^{\star}$$
 where $Q = \exp \begin{pmatrix} 0 & -1 & -1 & -1 \\ 1 & 0 & -1 & -1 \\ 1 & 1 & 0 & -1 & -1 \\ 1 & 1 & 1 & 0 & -1 \\ 1 & 1 & 1 & 1 & 0 \end{pmatrix}$

•
$$G_{\mathsf{Br}}$$
 is defined from $A = \begin{pmatrix} 5 & 0 & 0 & 0 & 0 \\ 0 & 4 & 0 & 0 & 0 \\ 0 & 0 & 3 & 0 & 0 \\ 0 & 0 & 0 & 2 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix}$

- The numerical solutions are computed using adaptive 4th-order Runge-Kutta scheme.
- Test the three flows presented before.

Brockett's choice

 $H(t) \to H_\infty = {\rm diag}([25,16,9,4,1])$ Diagonal terms are sorted in a descending order, w.r.t to A

$$\operatorname{spec} \mathsf{d} F_{H_{\infty}} = \begin{pmatrix} 0 & -9 & -32 & -63 & -96 \\ -9 & 0 & -7 & -24 & -45 \\ -32 & -7 & 0 & -5 & -16 \\ -63 & -24 & -5 & 0 & -3 \\ -96 & -45 & -16 & -3 & 0 \end{pmatrix}$$



Toda's choice

$$\operatorname{spec} \mathsf{d} F_{H_{\infty}} = \begin{pmatrix} 0 & -9 & -16 & -21 & -24 \\ -9 & 0 & -7 & -12 & -15 \\ -16 & -7 & 0 & -5 & -8 \\ -21 & -12 & -5 & 0 & -3 \\ -24 & -15 & -8 & -3 & 0 \end{pmatrix}$$



Wegner's choice

 $H(t) \rightarrow H_\infty = \mathsf{diag}([4,9,16,25,1])$

$$\mathsf{d}F_{H_{\infty}}(E_{i,j}) = -(\lambda_i - \lambda_j)^2 E_{i,j} \,.$$

$$\mathsf{spec} \, \mathsf{d}F_{H_{\infty}} = \begin{pmatrix} 0 & -25 & -144 & -441 & -9 \\ -25 & 0 & -49 & -256 & -64 \\ -144 & -49 & 0 & -81 & -225 \\ -441 & -256 & -81 & 0 & -576 \\ -9 & -64 & -225 & -576 & 0 \end{pmatrix}$$



Brockett's choice - other stable manifold

Other initial data :
$$\tilde{H}_0 = \tilde{Q} \begin{pmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 9 & 0 & 0 \\ 0 & 0 & 0 & 16 & 0 \\ 0 & 0 & 0 & 0 & 25 \end{pmatrix} \tilde{Q}^{\star}$$
, where $Q = \exp \begin{pmatrix} 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & -1 & -1 & 0 \\ 0 & 1 & 0 & -1 & -1 \\ 0 & 1 & 1 & 0 & -1 \\ 0 & 1 & 1 & 1 & 0 \end{pmatrix}$.

•
$$\tilde{H}(t) = \begin{pmatrix} 1 & (0) \\ (0) & K(t) \end{pmatrix}$$
 with spec $(K(t)) = \{4, 9, 16, 25\}.$

- The (floating-point arithmetic) numerical method preserves this particular subspace
- $H(t) \to H_{\infty} = \text{diag}([1, 25, 16, 9, 4])$ with exponential rate $\mathcal{O}(e^{-5t})$

$$\operatorname{spec} \mathsf{d}F_{H_{\infty}} = \begin{pmatrix} 0 & 24 & 30 & 24 & 12 \\ 24 & 0 & -9 & -32 & -63 \\ 30 & -9 & 0 & -7 & -24 \\ 24 & -32 & -7 & 0 & -5 \\ 12 & -63 & -24 & -5 & 0 \end{pmatrix}$$

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Grub's up !

À table !

Ăn trưa được phục vụ !