

Equazioni lineari matriciali proprietà, metodi numerici ed applicazioni

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• Sylvester matrix equation

$$A\mathbf{X} + \mathbf{X}B + D = 0$$

Eigenvalue problems, Control, Model Order Reduction, Assignment problems, Riccati equation

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• Lyapunov matrix equation

$$A\mathbf{X} + \mathbf{X}A^{\top} + D = 0, \qquad D = D^{\top}$$

Stability analysis in Control and Dynamical systems, Signal processing, eigenvalue computations

• Algebraic Riccati equation

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Lancaster-Rodman '95, Konstantinov-Gu-Mehrmann-Petkov, '02, Bini-Iannazzo-Meini '12

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• Multiterm linear matrix equation

$$A_1 \mathbf{X} B_1 + A_2 \mathbf{X} B_2 + \ldots + A_\ell \mathbf{X} B_\ell = C$$

Elliptic PDEs, PDEs with stochastic inputs, bilinear dynamical systems, etc.

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Focus: All or some of the matrices are large (and possibly sparse)

The Lyapunov equation.

$$A\mathbf{X} + \mathbf{X}A^{\top} + D = 0, \qquad A \text{ stable}$$

A	X	+	x	$A^{ op}$	+	D	= 0
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Example: For D = I and A symmetric, it holds that $\mathbf{X} = -\frac{1}{2}A^{-1}$

The Lyapunov equation. Some characterizations $AX + XA^{\top} + BB^{\top} = 0, \qquad A \in \mathbb{R}^{n \times n}$ stable

• The Applied Mathematician perspective

 ${f X}$ holds stability information of time-invariant dynamical system:

$$\mathbf{x}'(t) = A\mathbf{x}(t) + B\mathbf{u}(t), \qquad \mathbf{x}(0) = x_0$$

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• The Analyst perspective. Closed form solution:

$$X = -\frac{1}{2\pi} \int_{-\infty}^{\infty} (\imath \omega I - A)^{-1} B B^{\top} (\imath \omega I - A)^{-*} \mathrm{d}\omega = \int_{-\infty}^{0} e^{At} B B^{\top} e^{At} dt$$

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• The Algebraist perspective. Kronecker formulation:

$$(A \otimes I + I \otimes A)\mathbf{x} = b$$
 $\mathbf{x} = \operatorname{vec}(\mathbf{X}), \ b = \operatorname{vec}(BB^T)$

with $\mathcal{S} := A \otimes I + I \otimes A \in \mathbb{R}^{n^2 \times n^2}$

Linear systems vs linear matrix equations

Large linear systems:

 $S\mathbf{x} = b$,

- Krylov subspace methods (CG, MINRES, GMRES, BiCGSTAB, etc.)
- \bullet Preconditioners: find P such that

$$\mathcal{S}P^{-1}\widetilde{\mathbf{x}} = b \qquad \mathbf{x} = P^{-1}\widetilde{\mathbf{x}}$$

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Large linear matrix equations:

$$A\mathbf{X} + \mathbf{X}A^{\top} + BB^{\top} = 0$$

- No preconditioning to preserve symmetry
- X is a large, dense matrix \Rightarrow low rank approximation

$$\mathbf{X} \approx \widetilde{\mathbf{X}} = Z Z^{\top}, \quad Z \text{ tall}$$

The Kronecker sum matrix

 $\mathcal{S} := A \otimes I_n + I_n \otimes A,$

with \boldsymbol{A} symmetric and positive definite, banded with bandwidth \boldsymbol{b}

- Quantum Chemistry and Quantum dynamics
- Signal processing
- Numerical analysis
 - PDE discretizations: e.g., in Finite Differences, Finite Elements, Legendre Spectral Methods, Isogeometric Analysis, ...
- Multivariate Statistics

Sparsity and quasi-sparsity pattern properties of

 $f(\mathcal{S})$

 $f \in \{z^{-1}, e^z, z^{\frac{1}{2}}, \ldots\}$

Discretization of 2D Laplace operator on the unit square

$$\mathcal{S} := A \otimes I_n + I_n \otimes A, \qquad A = \operatorname{tridiag}(-1, 2, -1)$$

Sparsity pattern:



 $\mathsf{Matrix}\; \mathcal{S}$



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Sparsity pattern:



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The exponential decay of the entries of \mathcal{S}^{-1}

The classical bound (Demko, Moss & Smith):

If S spd is banded with bandwidth b, then

 $|(\mathcal{S}^{-1})_{ij}| \le \gamma q^{\frac{|i-j|}{b}}$

where

$$\begin{split} &\kappa = \lambda_{\max}(\mathcal{S})/\lambda_{\min}(\mathcal{S}) \text{ (cond. number of } \mathcal{S}) \\ &q := \frac{\sqrt{\kappa} - 1}{\sqrt{\kappa} + 1} < 1 \\ &\gamma := \max\{\lambda_{\min}(\mathcal{S})^{-1}, \hat{\gamma}\}, \text{ and } \hat{\gamma} = \frac{(1 + \sqrt{\kappa})^2}{2\lambda_{\max}(\mathcal{S})} \end{split}$$

 $(\lambda_{\min}(\mathcal{S}), \lambda_{\max}(\mathcal{S}) \text{ smallest and largest eigenvalues of } \mathcal{S})$

Many contributions: Bebendorf, Hackbusch, Benzi, Boito, Razouk, Golub, Tuma, Concus, Meurant, Mastronardi, Ng, Tyrtyshnikov, Nabben, ...

The actual decay



... a very peculiar pattern \Rightarrow much higher sparsity

Where do the repeated peaks come from?

For $S = A \otimes I_n + I_n \otimes A \in \mathbb{R}^{n^2 \times n^2}$:

$$x_t := (\mathcal{S}^{-1})_{:,t} = \mathcal{S}^{-1}e_t \qquad \Leftrightarrow \qquad \text{Solve}: \ \mathcal{S}\mathbf{x}_t = e_t$$

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Let

$$\mathbf{X}_t \in \mathbb{R}^{n \times n}$$
 be such that $\mathbf{x}_t = \operatorname{vec}(\mathbf{X}_t)$
 $E_t \in \mathbb{R}^{n \times n}$ be such that $e_t = \operatorname{vec}(E_t)$
Then

$$\mathcal{S}\mathbf{x}_t = e_t \qquad \Leftrightarrow \qquad A\mathbf{X}_t + \mathbf{X}_t A = E_t$$

The Poisson equation - revisited

 $-u_{xx} - u_{yy} = f$, in $\Omega = (0, 1)^2$

+ Dirichlet b.c. (zero b.c. for simplicity)



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FD Discretization: $U_{i,j} \approx u_{x_i,y_j}$, with (x_i, y_j) interior nodes, so that

$$u_{xx}(x_i, y_j) \approx \frac{U_{i-1,j} - 2U_{i,j} + U_{i+1,j}}{h^2} = \frac{1}{h^2} [1, -2, 1] \begin{bmatrix} U_{i-1,j} \\ U_{i,j} \\ U_{i+1,j} \end{bmatrix}$$
$$u_{yy}(x_i, y_j) \approx \frac{U_{i,j-1} - 2U_{i,j} + U_{i,j+1}}{h^2} = \frac{1}{h^2} [U_{i,j-1}, U_{i,j}, U_{i,j+1}] \begin{bmatrix} 1 \\ -2 \\ 1 \end{bmatrix}$$

 $A\mathbf{U} + \mathbf{U}A = F, \quad F_{ij} = f(x_i, y_j)$

For S the 2D Laplace operator, $t = 1, ..., n^2$ t = 35, $S\mathbf{x}_t = e_t \iff A\mathbf{X}_t + \mathbf{X}_t A = E_t$



matrix E_t

matrix \mathbf{X}_t

 and

For S the 2D Laplace operator, $t = 1, ..., n^2$ t = 35, $S\mathbf{x}_t = e_t \iff A\mathbf{X}_t + \mathbf{X}_t A = E_t$



matrix E_t and matrix \mathbf{X}_t E_t has only one nonzero element Lexicographic order: $(E_t)_{ij}$, $j = \lfloor (t-1)/n \rfloor + 1$, $i = tn \lfloor (t-1)/n \rfloor$



Left: Row of S^{-1} Right: same row on the grid



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Left: Row of \mathcal{S}^{-1} Right: same row on the grid



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Qualitative bounds (more general than for the Laplacian!) Let $\kappa = \lambda_{\max}/\lambda_{\min} = \operatorname{cond}(A)$ i)Assume $\ell, i, m, j : \ell \neq i, m \neq j$. $\mathfrak{n}_2 := |\ell - i| + |m - j| - 2 > 0$ $|(S^{-1})_{k,t}| \leq \frac{\sqrt{\kappa^2 + 1}}{2\lambda_{\min}} \frac{1}{\sqrt{\mathfrak{n}_2}}.$

ii)Assume ℓ, i, m, j : $\ell = i$ or m = j. $\mathfrak{n}_1 := |\ell - i| + |m - j| - 1 > 0$



Example. Legendre stiffness matrix (scaled to have unit peak)

 $A = \operatorname{tridiag}(\delta_k, \underline{\gamma_k}, \delta_k)$



$$\gamma_k = \frac{2}{(4k-3)(4k+1)}$$

 $k = 1, \dots, n, \text{ and}$
 $\delta_k = \frac{-1}{(4k+1)\sqrt{(4k-1)(4k+3)}}$
 $k = 1, \dots, n-1$

Canuto, Simoncini, Verani 2014

Connections to point-wise estimates for discrete Laplacian

For the discrete Green function G_h on the discrete *d*-dimensional grid R_h , there exist constants h_0 and C such that for $h \leq h_0$, $x, y \in R_h$,

$$G_h(x,y) \le \begin{cases} C \log \frac{C}{|x-y|+h} & \text{if } d = 2\\ \frac{C}{(|x-y|+h)^{d-2}} & \text{if } d \ge 3 \end{cases}$$

(Bramble & Thomee, '69)

Our estimate: entries depend on inverse square root of the distance!

Typical decay plot for $f(\mathcal{S})$, with \mathcal{S} Laplace operator as before



$f(\mathcal{S}) = e^{-5\mathcal{S}}$	$f(\mathcal{S}) = \mathcal{S}^{-\frac{1}{2}}$
J(C) C	$J(\mathbf{c}) \mathbf{c}$

(Bounds for Laplace or Stieltjes functions)

In general, $\mathcal{S}=A_1\oplus A_2:=A_1\otimes I+I\otimes A_2$, A_1,A_2 banded spd Benzi, Simoncini 2015

Generalizations

- Three-dimensional case
- (banded) Non-symmetric matrices
- "Quasi" Kronecker structure
- Numerical solution of PDEs on structured grids

 $-\Delta u = 1, \quad \Omega = (0,1)^3 \quad \Rightarrow \quad \mathcal{S} = (A \otimes I \otimes I + I \otimes A \otimes I + I \otimes I \otimes A)$ CG for $\mathcal{S}x = b$ vs Iterative solver for $(I \otimes A + A \otimes I)\mathbf{U} + \mathbf{U}A = F$ $A \in \mathbb{R}^{n \times n}, \quad \mathcal{S} \in \mathbb{R}^{n^3 \times n^3}, \qquad n = 50$

 $-\Delta u = 1, \quad \Omega = (0,1)^3 \quad \Rightarrow \quad \mathcal{S} = (A \otimes I \otimes I + I \otimes A \otimes I + I \otimes I \otimes A)$ CG for Sx = b vs Iterative solver for $(I \otimes A + A \otimes I)\mathbf{U} + \mathbf{U}A = F$ $A \in \mathbb{R}^{n \times n}$, $S \in \mathbb{R}^{n^3 \times n^3}$, n = 5010⁰ CG 10⁻¹ 10⁻² residual norm 10-3 10-2 PCG 10⁻⁶ 10⁻⁷ Sylv 10^{-8} 10⁻⁹ 10 50 80 20 30 40 60 70 number of iterations/ space dim PCG CG Matrix Eqn solver Comput. Time 2.91 0.56 0.08

Generalizations. Solutions to PDEs



In polar coordinates (r, θ) : $-u_{rr} - \frac{1}{r}u_r - u_{\theta\theta} = \tilde{f}$ $\Rightarrow \qquad A_1 \mathbf{X} + \mathbf{X}A_2 = F$

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In polar coordinates (r, θ) : $-u_{rr} - \frac{1}{r}u_r - u_{\theta\theta} = \tilde{f}$ $\Rightarrow \qquad A_1 \mathbf{X} + \mathbf{X}A_2 = F$

Structured grids

Applications

- Computational Aero- and Fluid-Dynamics
- Seminconductor devices
- Object modelling
- Parallel computation
- ...

Classical strategies (building blocks)

- Conformal mappings (Boundary-fitted curvilinear coord.)
- Algebraic grid generators (Transfinite interpolation)
- Elliptic, hyperbolic grids with controls
- Variational methods
- ...

Grid generation. An example



(grids from http://www.math.fsu.edu/ okhanmoh/research.html)

Conclusions

- Matrix equations have very broad applicability (structure recurrent in many application problems...)
- Recent appropriate computational devices
- Important tool for matrix sparsity analysis

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