Some more comments on the normal equations: With a focus on discretisation of partial differential equations



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Due Giorni Di Algebra Lineare Numerica, 20th January 2025









Let us consider the following linear system of equations

$$\underline{\underline{A}}\underline{x} = \underline{b}, \qquad \underline{\underline{A}} \in \mathbb{R}^{n \times n}, \quad \underline{x}, \underline{b} \in \mathbb{R}^{n}.$$

$$A^T \neq A$$

In order to solve the system, we can consider the normal equation, i.e.

$$B := \underline{\underline{A}}^T \underline{\underline{A}} \underline{x} = \underline{\underline{A}}^T \underline{\underline{b}}$$



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- Unfortunately the condition number of  $\underline{\underline{A}}^T\underline{\underline{A}}$  is the square of the condition number of  $\underline{A}$ .
- We now have a symmetric positive definite system, that can be solved using CG (CGNE).

## HOW CAN WE PRECONDITION THE NORMAL EQUATIONS?





SIREV Vol. 64, Iss. 3, 2022 (A. Wathen),

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 $\underline{\underline{P}}$  is a good preconditioner if  $\underline{\underline{P}}^{-1}\underline{\underline{A}}$  has clustered eigenvalues.

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$$A = \begin{bmatrix} b_0 & & & & \\ & \ddots & & \\ & & b_{n-1} \end{bmatrix}, \qquad P = \begin{bmatrix} & & b_0 \\ & \ddots & \\ b_{n-1} & & \end{bmatrix}.$$







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$$P^{-1}A = \begin{bmatrix} & & 1 \\ & \ddots & \\ 1 & & \end{bmatrix}, \qquad G^{-1}B = \begin{bmatrix} (b_0/b_{n-1})^2 & & & \\ & \ddots & & \\ & & (b_{n-1}/b_0)^2 \end{bmatrix}.$$







SIREV Vol. 64, Iss. 3, 2022 (A. Wathen), QJRMS Vol. 64, Iss. 3, 2022 (S. Gratton, Et Al.).

# Gratton-Gürol-Simon-Toint

If the matrix P is such that  $\|I - AP^{-1}\|_2 \le \sqrt{2} - 1 - \delta$ , then  $\Lambda(G^{-1}B) \subset (\sqrt{2}\delta + \delta^2, 2 - \sqrt{2}\delta - \delta^2).$ 







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We consider the matrix  $T := I - AP^{-1}$ , and expand  $G^{-1}B$  as

$$G^{-1}B = P^{-1}P^{-T}A^{T}A \sim P^{-T}A^{T}AP^{-1} = I - T - T^{T} + T^{T}T.$$

Since  $\Lambda(G^{-1}B) \subset [-\|G^{-1}B\|_2, \|G^{-1}B\|_2]$ , we can easily see that

$$-1 - 2||T||_2 - ||T||_2^2 \le \lambda \le 1 + 2||T||_2 + ||T||_2^2$$
.

Substituing  $||I - AP^{-1}||_2 \le \sqrt{2} - 1 - \delta$  we obtained the desired result.



We would like to give a different intuition of good preconditioners for normal equations. To this aim we consider the previously observed similarity,

$$G^{-1}B = P^{-1}P^{-T}A^{T}A \sim P^{-T}A^{T}AP^{-1} = (AP^{-1})^{T}(AP^{-1}).$$

Hence, the closer the matrix  $AP^{-1}$  is to an orthogonal matrix, the closer  $G^{-1}B$  is to the identity matrix.



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## **Cross preconditioning**

We say that the preconditioner P is a good **left** preconditioner for the normal equations if it is a good **right** preconditioner for  $\underline{\underline{A}}$ , in the sense that  $\underline{\underline{AP}}^{-1}$  has clustered singular values.



The ideal preconditioner for  $\underline{A}$  is unique, up to scaling, and it is the inverse of  $\underline{A}$ .



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There is a much wider choice of good cross preconditioners for the normal equations, in fact the space of orthogonal matrices has dimension n(n-1)/2.



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# QR decomposition

We can construct an ideal preconditioner using the QR decomposition of  $\underline{A}$ , i.e.

$$\underline{\underline{P}} = \underline{\underline{R}}, \ \underline{\underline{A}} = \underline{\underline{Q}}_{OR}\underline{\underline{R}}.$$



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# Polar decomposition

We can construct an ideal preconditioner using the polar decomposition of  $\underline{A}$ , i.e.

$$\underline{\underline{P}} = (\underline{\underline{A}}^T \underline{\underline{A}})^{\frac{1}{2}}, \ \underline{\underline{A}} = \underline{\underline{Q}}_{\underline{P}}\underline{\underline{P}}.$$

### ADVECTION DIFFUSION ODE - CROSS PRECONDITIONING



We consider the classical advection-diffusion ODE in one dimension, i.e.

$$-\nu\ddot{u}+\beta\dot{u}=f \text{ in } (a,b)\subset\mathbb{R},$$
  
$$u(a)=0,\ u(b)=1,\ \nu,\beta\in\mathbb{R}_{\geq 0}.$$



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For the moment we will consider neither diffusion nor advection-dominated regimes, i.e.  $\nu \approx \beta$ , and discretisation over an equi-spaced mesh of step-size h. Such a discretisation results in the matrix

$$\underline{\underline{A}} = \operatorname{tridiag}\left(-\frac{\nu}{h^2} - \frac{\beta}{2h}, \frac{2\nu}{h^2}, -\frac{\nu}{h^2} + \frac{\beta}{2h}\right)$$



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$$\underline{\underline{A}} = \operatorname{tridiag} \left( -\frac{\nu}{\mathit{h}^2} - \frac{\beta}{2\mathit{h}}, \frac{2\nu}{\mathit{h}^2}, -\frac{\nu}{\mathit{h}^2} + \frac{\beta}{2\mathit{h}} \right)$$



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	n	QR	RQ	$Q(A^TA)^{1/2}$	$(AA^T)^{1/2}Q$
1	0	2	12	2	4
1	00	2	-	2	6
10	000	2 2 2	-	2	7

Table: Comparison of the number of iterations for different preconditioners for the left preconditioned normal equation. The CGNE method was terminated when the absolute residual was less than  $10^{-12}$ . If the method did not converge in 1000 iterations, we marked the number of iterations with a dash.

### ADVECTION DIFFUSION PDE



We consider the classical advection-diffusion PDE in two dimensions, i.e.

$$\mathcal{L}u := -\nu\Delta u + \underline{\beta} \cdot \nabla u = f \text{ in } \Omega \subset \mathbb{R}^d,$$
 
$$u = g \text{ on } \partial\Omega, \text{ with } \nu \ll \|\beta\|, \ \nabla \cdot \beta = 0.$$



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## Finite Element Discretisation

Fix a discrete space  $V_h \subset H^1_0(\Omega)$  and look for  $u_h \in V_h$  such that

$$(\hat{\mathcal{L}}u_h,v_h)=\nu(\nabla u_h,\nabla v_h)_{L^2(\Omega)}+(\beta\cdot\nabla u_h,v_h)_{L^2(\Omega)}=(f,v_h)_{L^2(\Omega)} \text{ for any } v_h\in V_h.$$



We now need to understand what are the normal equations associated with the linear system,

$$A\underline{x} = \underline{b}$$
, with  $A_{ij} = (\hat{\mathcal{L}}\varphi_i, \varphi_j)_{L^2(\Omega)}$  and  $b_j = (f, \varphi_j)_{L^2(\Omega)}$ .



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The first thing we need to understand is what is  $\underline{\underline{A}}^T$ , in fact  $\underline{\underline{A}}^T$  is neither **Hilbert adjoint** of A nor the **Banach adjoint** seen as the operator  $A: V_h \subset H_0^1(\Omega) \to H^{-1}(\Omega) \subset V_h'$ .



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In fact,  $A^T$  is an operator itself of the form  $A^T: V_h \subset H^1_0(\Omega) \to H^{-1}(\Omega) \subset V_h'$  which corresponds to the discretisation of the **Hilbert adjoint** of  $\mathcal{L}$ , i.e.

$$A_{ij}^T = A_{ji} = (\hat{\mathcal{L}}\varphi_j, \varphi_i)_{L^2(\Omega)} = (\varphi_j, \hat{\mathcal{L}}^*\varphi_i)_{L^2(\Omega)} = (\hat{\mathcal{L}}^*\varphi_i, \varphi_j)_{L^2(\Omega)},$$

### THE NORMAL EQUATIONS - PRIMAL DUAL ERROR



If we consider the classical normal equations, i.e.  $A^TAx = A^Tb$ .

### **Primal Dual Error**

We notice that there is a primal dual error in the classical formulation of the normal equations.

$$V_h \subset H^1_0(\Omega) \stackrel{A}{\longrightarrow} H^{-1} \subset V'_h$$

$$V_h \subset H^1_0(\Omega) \xrightarrow{A} H^{-1} \subset V_h'$$
  $V_h \subset H^1_0(\Omega) \xrightarrow{A^T} H^{-1} \subset V_h'$ 

### THE NORMAL EQUATIONS - PRIMAL DUAL ERROR



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To make sense of the normal equations we need to consider a Riesz map  $T:V_h'\to V_h$ .

$$V_h \subset H^1_0(\Omega) \xrightarrow{A} H^{-1} \subset V'_h \xrightarrow{T} V_h \subset H^1_0(\Omega) \xrightarrow{A^T} H^{-1} \subset V'_h$$



The Riesz map gives rise to a discrete operator  $T: V_h' \to V_h$ , which is **symmetric and positive definite**. Therefore if we consider the normal equations with respect to the Riesz map, i.e.

$$\underline{\underline{A}}^T T \underline{\underline{A}} \underline{x} = \underline{\underline{A}}^T T \underline{b},$$

we can rewrite them using a Cholesky factorisation of T, i.e.  $T = C^T C$ .

$$(CA)^T(CA)\underline{x} = (CA)^TC\underline{b},$$

hence the previous normal equation are associated with the linear system  $CA\underline{x} = C\underline{b}$ .

- The normal equations are still symmetric and positive definite. Hence we can solve them using CGNE. The cross-preconditioning idea is still applicable.
- The condition number of the normal equations is the square of the condition number of the original system.

# THE NORMAL EQUATIONS – $L^2$ -RIESZ MAP



We can consider as Riesz map the  $L^2$ -Riesz map, i.e.

$$(\mathit{Tf}, v_h)_{L^2(\Omega)} = \langle f, v_h 
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Using the  $L^2$ -Riesz map the new normal is approximating, in the limit  $\nu \to 0$ , the problem: find  $u \in H^1_0(\Omega)$  such that

$$(\beta \otimes \beta \nabla u, \nabla v)_{L^2(\Omega)} = (g, v)_{L^2(\Omega)}$$
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$\nu$	CGNE Iterations		
$1 \cdot 10^{-2}$	4231		
$5 \cdot 10^{-3}$	3803		
$2.5 \cdot 10^{-3}$	3327		
$1.25\cdot 10^{-3}$	2419		

Table: The CGNE methods were terminated when the absolute residual was less than  $10^{-5}$ .

$$(\beta \otimes \beta \nabla u, \nabla v)_{L^2(\Omega)} = (g, v)_{L^2(\Omega)}$$
 for any  $v \in H^1_0(\Omega)$ .

Due to the function space involved in the weak form, we chose the wrong Riesz map.

$$H_0^1(\Omega) \longrightarrow H^{-1} \subset L^{2'} \stackrel{T^{-1}}{\longrightarrow} L^2 \not\subset H_0^1(\Omega) \longrightarrow H^{-1}$$

## THE NORMAL EQUATIONS – $H^1$ -RIESZ MAP



We can consider as Riesz map the  $H^1$ -Riesz map, i.e.

$$(\nabla Tf, \nabla v_h)_{L^2(\Omega)} = \nu^{-1} \langle f, v_h \rangle, \ \forall v_h \in V_h, f \in V_h'.$$

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Using this Riesz map the normal equations  $\underline{\underline{A}}^T T \underline{\underline{A}} \underline{x} = \underline{\underline{A}}^T T \underline{\underline{b}}$  is approximating the problem: find  $u \in H_0^1(\Omega)$  such that

$$\nu(\nabla u, \nabla v)_{L^2(\Omega)} + \nu^{-1}(\Pi_{\nabla}\beta u, \Pi_{\nabla}\beta v)_{L^2(\Omega)}, \text{ for any } v \in H^1_0(\Omega).$$

# THE NORMAL EQUATIONS – H¹-RIESZ MAP



We can consider as Riesz map the  $H^1$ -Riesz map, i.e.

$$(\nabla Tf, \nabla v_h)_{L^2(\Omega)} = \nu^{-1} \langle f, v_h \rangle, \ \forall v_h \in V_h, f \in V_h'.$$

ν	32 × 32	64 × 64	128 × 128
$1 \cdot 10^{-2}$	2	2	2
$5 \cdot 10^{-3}$	3	3	3
$2.5 \cdot 10^{-3}$	3	3	3
$1.25\cdot 10^{-3}$	3	3	3

Table: The CGNE methods were terminated when the absolute residual was less than  $10^{-5}$ .

Using this Riesz map the normal equations  $\underline{\underline{A}}^T T \underline{\underline{A}} \underline{x} = \underline{\underline{A}}^T T \underline{\underline{b}}$  is approximating the problem: find  $u \in H_0^1(\Omega)$  such that

$$\nu(\nabla u, \nabla v)_{L^2(\Omega)} + \nu^{-1}(\Pi_{\nabla}\beta u, \Pi_{\nabla}\beta v)_{L^2(\Omega)}, \text{ for any } v \in H^1_0(\Omega).$$





Find  $u_h \in V_h$  such that  $\nu^{-1}(\beta u_h, \beta v_h)_{L^2(\Omega)}$ , for any  $v_h \in V_h$ .

ν	32 × 32	64 × 64	$128 \times 128$	$256 \times 256$	512 × 512
$1\cdot 10^{-2}$	9	14	21	24	26
$5\cdot 10^{-3}$	13	13	19	28	33
$2.5 \cdot 10^{-3}$	19	17	17	25	37
$1.25\cdot 10^{-3}$	27	24	21	22	33

Table: Comparison of the number of iterations for the CGNE method preconditioned by the inversion via PETSc GAMG, for different values of  $\nu$  and different mesh sizes. The wind is fixed to  $\beta=(1,0)$  and as right-hand side we consider the function  $f(x,y)\equiv 1$ . The CGNE method was terminated when the absolute residual was less than  $10^{-5}$ .

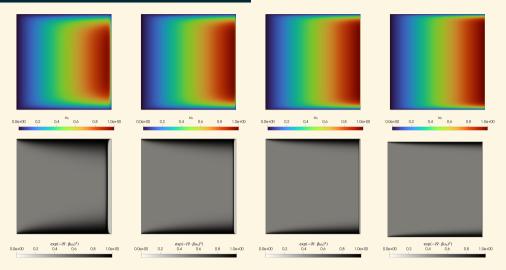


Figure: The discrete solution  $u_h$  of the advection-diffusion equation (1) for different value of  $\nu$  at the finest mesh size 512  $\times$  512, together with  $exp(-|\nabla \cdot \beta u_h|^2)$ .





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ν	32 × 32	64 × 64	128 × 128	256 × 256
$1\cdot 10^{-2}$	10	15	20	23
$5\cdot 10^{-3}$	11	15	22	30
$2.5 \cdot 10^{-3}$	17	16	21	32
$1.25 \cdot 10^{-3}$	26	24	23	30

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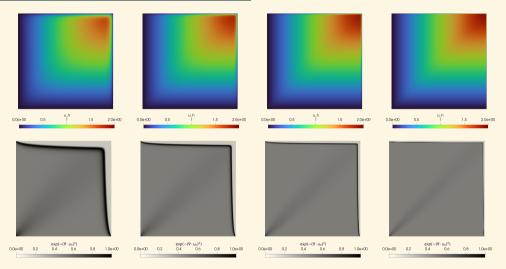


Figure: The discrete solution  $u_h$  of the convection-diffusion equation (1), with  $\sqrt{2}\beta = (1,1)$ , for different values of  $\nu$  at the finest mesh size 512  $\times$  512, together with  $\exp(-|\nabla \cdot \beta u_h|^2)$ .





Find  $u_h \in V_h$  such that  $\nu^{-1}(\Pi_{\nabla}\beta u_h, \Pi_{\nabla}\beta v_h)_{L^2(\Omega)}$ , for any  $v_h \in V_h$ .

ν	32 × 32	64 × 64	128 × 128
$1\cdot 10^{-2}$	14	22	40
$5\cdot 10^{-3}$	16	21	33
$2.5 \cdot 10^{-3}$	22	22	29
$1.25\cdot 10^{-3}$	30	30	34

Table: Comparison of the number of iterations for the CGNE method preconditioned by symmetric successive over-relaxation, for different values of  $\nu$  and different mesh sizes. The wind is fixed to  $\sqrt{2}\beta=(1,1)$  and as right-hand side we consider the function  $f(x,y)\equiv 1$ . The CGNE method was terminated when the absolute residual was less than  $10^{-5}$ .





Find  $u_h \in V_h$  such that  $\nu^{-1}(\Pi_{\nabla}\beta u_h, \Pi_{\nabla}\beta v_h)_{L^2(\Omega)}$ , for any  $v_h \in V_h$ .

ν	32 × 32	64 × 64	128 × 128
$1\cdot 10^{-2}$	4	5	8
$5\cdot 10^{-3}$	4	5	7
$2.5 \cdot 10^{-3}$	5	5	7
$1.25 \cdot 10^{-3}$	7	7	7

Table: Comparison of the number of iterations for the CGNE method preconditioned by geometric multigird with SOR smoothing, for different values of  $\nu$  and different mesh sizes. The wind is fixed to  $\sqrt{2}\beta=(1,1)$  and as right-hand side we consider the function  $f(x,y)\equiv 1$ . The CGNE method was terminated when the absolute residual was less than  $10^{-5}$ .



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- Apply **normal preconditioning** to other PDEs such as the Helmholtz equation, using as Riesz map the T-coercive map. We would also like to study the Oseen and Helmholtz equation and  $C^1$  nearly singular problems such as the Helmholtz–Korteweg equation.



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- Understand how to efficiently compute the polar decomposition so that we can construct a good **cross preconditioner** starting from the normal PDE, for LSQR type methods.

# THANK YOU! Lorenzo now accepts questions.

Some more comments on the normal equations: With a focus on discretisation of partial differential equations

L. Lazzarino, Y. Nakatsukasa, Umberto Zerbinati\*