

# ON BEST CONDITIONED MATRICES<sup>1</sup>

G. E. FORSYTHE AND E. G. STRAUS

1. **Main theorems.** Let  $A$  be a positive definite Hermitian matrix of finite order, and let  $\Lambda$  and  $\lambda$  be its maximal and minimal eigenvalue respectively. The *condition number* of  $A$  is the ratio  $P(A) = \Lambda/\lambda$  introduced by Todd [1]. Let  $\mathcal{T}$  be a class of regular linear transformations. Define  $A^T = T^*AT$ . We say that  $A$  is *best conditioned with respect to*  $\mathcal{T}$  if  $P(A^T) \geq P(A)$  for all  $T \in \mathcal{T}$ .

In order to investigate whether  $A$  is best conditioned we remember that

$$(1) \quad \Lambda = \max_x \frac{x^*Ax}{x^*x}, \quad \lambda = \min_x \frac{x^*Ax}{x^*x}$$

and hence

$$(2) \quad P(A) = \max_{\|x\|=\|y\|=1} \frac{x^*Ax}{y^*Ay}.$$

We introduce the abbreviation  $R = R(T) = (T^*)^{-1}T^{-1}$ . Now let  $\Lambda^T, \lambda^T$  be the extremal eigenvalues of  $A^T$ . Setting  $u = Tx$ , we obtain from (1) and (2):

$$\begin{aligned} \Lambda^T &= \max_x \frac{x^*T^*ATx}{x^*x} = \max_u \frac{u^*Au}{u^*Ru}, \\ \lambda^T &= \min_x \frac{x^*T^*ATx}{x^*x} = \min_u \frac{u^*Au}{u^*Ru}, \\ (3) \quad P(A^T) &= \max_{\|u\|=\|v\|=1} \frac{u^*Au}{v^*Av} \cdot \frac{v^*Rv}{u^*Ru}. \end{aligned}$$

Thus, if we let  $S_\Lambda, S_\lambda$  be the sets of unit eigenvectors of  $A$  belonging to  $\Lambda$  and  $\lambda$  respectively, then we obtain from (2)

$$(4) \quad P(A) = \frac{x^*Ax}{y^*Ay}, \quad x \in S_\Lambda, y \in S_\lambda.$$

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Hence from (3) and (4)

$$P(A^T) \geq P(A) \max_{u \in S_\Lambda, v \in S_\lambda} \frac{v^* R v}{u^* R u}.$$

We thus have proved:

LEMMA. If  $\max_{u \in S_\Lambda, v \in S_\lambda} (v^* R v / u^* R u) \geq 1$  for all  $T \in \mathcal{T}$ , then  $A$  is best conditioned with respect to  $\mathcal{T}$ .

It will be convenient to introduce the concept of "separability by  $\mathcal{T}$ ":

DEFINITION. The sets  $S_1, S_2$  are separable by  $\mathcal{T}$  if there exists a  $T \in \mathcal{T}$  and a constant  $k$  so that

$$x^* R x < k < y^* R y,$$

for all  $x$  in one  $S_j$  and all  $y$  in the other.

Obviously, if  $S_1, S_2$  are not separable by  $\mathcal{T}$ , then

$$(5) \quad \sup_{x \in S_1, y \in S_2} \frac{x^* R x}{y^* R y} \geq 1 \quad \text{for all } T \in \mathcal{T}.$$

Combining (5) with the lemma, we have proved

THEOREM 1. If  $S_\Lambda$  and  $S_\lambda$  are not separable by  $\mathcal{T}$ , then  $A$  is best conditioned with respect to  $\mathcal{T}$ .

The converse to Theorem 1 is not true without further conditions on  $\mathcal{T}$ . As such a condition we introduce the following concept:

DEFINITION. A set  $\mathcal{T}$  of regular linear transformations is called *infinitesimally complete* if, for every  $T \in \mathcal{T}$ , there exist arbitrarily small positive  $\epsilon, \epsilon'$  such that there are  $T_\epsilon, T_{\epsilon'} \in \mathcal{T}$  with

$$I + \epsilon R = c(T_\epsilon^*)^{-1} T_\epsilon^{-1}, \quad I - \epsilon' R = c'(T_{\epsilon'}^*)^{-1} T_{\epsilon'}^{-1},$$

where  $c, c'$  are (positive) numbers.

THEOREM 2. If  $\mathcal{T}$  is infinitesimally complete and  $S_\lambda, S_\Lambda$  are separable by  $\mathcal{T}$ , then  $A$  is not best conditioned with respect to  $\mathcal{T}$ .

PROOF. By the hypothesis of separability there exists a  $T \in \mathcal{T}$  and a  $k > 0$  such that either

$$(I) \quad x^* R x > k > y^* R y \quad \text{for all } x \in S_\Lambda, y \in S_\lambda,$$

or

$$(II) \quad x^* R x < k < y^* R y \quad \text{for all } x \in S_\Lambda, y \in S_\lambda.$$

In case (I) we have  $y^*Ry/x^*Rx < 1$  for all  $x \in S_\Delta, y \in S_\lambda$ . Hence there exist neighborhoods  $U_\Delta, U_\lambda$  of  $S_\Delta, S_\lambda$  on the unit sphere  $S$  so that for every  $\epsilon > 0$

$$\sup_{z \in U_\Delta, v \in U_\lambda} \frac{y^*(I + \epsilon R)y}{x^*(I + \epsilon R)x} < 1.$$

Define  $U$  to be the Cartesian product  $U_\Delta \times U_\lambda$ . Then

$$(6) \quad \left( \max_{(x,y) \in U} \frac{x^*Ax}{y^*Ay} \right) \cdot \left( \sup_{(x,y) \in U} \frac{y^*(I + \epsilon R)y}{x^*(I + \epsilon R)x} \right) < P(A).$$

Let  $F = S \times S - U$ . Then  $\max_{(x,y) \in F} (x^*Ax/y^*Ay) < P(A)$ . Hence we may fix  $\epsilon$  so small that

$$(7) \quad \left( \max_{(x,y) \in F} \frac{x^*Ax}{y^*Ay} \right) \cdot \left( \max_{(x,y) \in F} \frac{y^*(I + \epsilon R)y}{x^*(I + \epsilon R)x} \right) < P(A).$$

By the infinitesimal completeness of  $\mathfrak{C}$ , there is a  $T_\epsilon \in \mathfrak{C}$  such that

$$(8) \quad \frac{y^*R_\epsilon y}{x^*R_\epsilon x} = \frac{y^*(I + \epsilon R)y}{x^*(I + \epsilon R)x}, \quad \text{where } R_\epsilon = R(T_\epsilon).$$

Putting (8) into (6) and (7), we then see from (3) that

$$(9) \quad P(AT_\epsilon) < P(A).$$

The proof of (9) in Case (II) is entirely analogous, if we replace  $I + \epsilon R$  in (6), (7), (8) by  $I - \epsilon'R$ . But (9) proves the theorem.

**2. Applications.** As examples of infinitesimally complete classes  $\mathfrak{C}$  we may cite:

(i) *Quasidiagonal* matrices. These are matrices of form

$$T = \begin{bmatrix} M_1 & 0 & \cdots & 0 \\ 0 & M_2 & \cdots & 0 \\ \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot \\ 0 & 0 & \cdots & M_s \end{bmatrix},$$

where each  $M_i$  is a square matrix of arbitrary preassigned order. An important subclass is the following:

(ii) *Diagonal* (or real diagonal or positive diagonal) matrices  $D$ . Here forming  $D^*AD$  is a special case of the common practice of preconditioning  $A$  by scaling rows and columns. This is used, for example, to make  $A$  more easily invertible by a numerical process. For numerical operations on a general nonsingular matrix  $C$  the condition

number  $P(A)$ , where  $A = CC^*$ , is often significant. Preconditioning of  $C$  by scaling rows alone yields a matrix  $C_1 = D^*C$ , for which  $C_1C_1^* = D^*AD$ . Minimizing  $P(D^*AD)$  (at least approximately) thus has practical importance for both Hermitian and general matrices.

If  $D$  is a regular diagonal matrix, then  $x^*Rx$  assumes the particularly simple form

$$(10) \quad x^*Rx = x^*(D^*)^{-1}D^{-1}x = \sum_{i=1}^n |d_{ii}|^{-2} |x_i|^2.$$

Thus separability by class (ii) means  $S_\Lambda$  and  $S_\lambda$  can be separated by an axis-oriented, origin-centered ellipsoid. From (10) we can establish

**THEOREM 3.** *A sufficient condition for  $A$  to be best conditioned with respect to class (ii) is that, for some pair of eigenvectors  $x^\Lambda, x^\lambda$  belonging to  $\Lambda, \lambda$ ,*

$$(11) \quad |x_i^\Lambda| = |x_i^\lambda| \quad (i = 1, \dots, n).$$

*Moreover, if  $\Lambda, \lambda$  are simple eigenvalues, (11) is also necessary.*

**PROOF.** If (11) holds, then, by (10),  $x^*Rx$  assumes the same value for both  $x^\Lambda$  and  $x^\lambda$ . The sufficiency then follows from Theorem 1. On the other hand, if  $\Lambda$  and  $\lambda$  are simple then  $S_\Lambda, S_\lambda$  consist of two points each. We then see that (11) is necessary and sufficient for separability of  $S_\Lambda$  and  $S_\lambda$ . This proves the necessity.

Note that (11) says that  $x^\Lambda, x^\lambda$  are reflections of each other in some coordinate subspace.

When  $\Lambda$  or  $\lambda$  is multiple, there are inseparable  $S_\Lambda, S_\lambda$  containing no  $x^\Lambda, x^\lambda$  which are reflections of each other. For an example of this, let

$$A = \begin{bmatrix} 1 & \alpha & \alpha \\ \alpha & 1 & \alpha \\ \alpha & \alpha & 1 \end{bmatrix}, \quad 0 < \alpha < 1.$$

Here  $\Lambda = 1 + 2\alpha$ ;  $S_\Lambda$  consists of two points  $\pm P$ , where  $P = (1, 1, 1)/3^{1/2}$ . Also  $\lambda = 1 - \alpha$  (double root), and  $S_\lambda$  is the circle  $x + y + z = 0, x^2 + y^2 + z^2 = 1$ . Now we show that it is impossible to separate  $S_\lambda$  from  $S_\Lambda$  by any quadratic surface  $f(x, y, z) = ax^2 + by^2 + cz^2 = d$ . First,  $f(P) = (a + b + c)/3$ . Let  $r = 1/2^{1/2}$ . Take, on  $S_\lambda$ ,  $P_1 = (r, -r, 0)$ ,  $P_2 = (r, 0, -r)$ , and  $P_3 = (0, r, -r)$ . Then  $f(P_1) = (a + b)/2$ ,  $f(P_2) = (a + c)/2$  and  $f(P_3) = (b + c)/2$ . Hence  $f(P) = [f(P_1) + f(P_2) + f(P_3)]/3$ , and  $f(P)$  must lie between the extreme values of the  $f(P_i)$ .

Theorem 3 will be applied to prove a conjecture of Young [2]. The conjecture is significant for an iterative solution of certain systems of linear equations.

THEOREM 4. *A positive definite Hermitian matrix of form*

$$(12) \quad Q = \begin{bmatrix} I_p & B \\ B^* & I_q \end{bmatrix},$$

where  $I_p, I_q$  are unit matrices, is always best conditioned with respect to class (ii).

PROOF. Let  $r$  be the rank of  $B$ . The semidefinite matrix  $B^*B$  has exactly  $r$  positive eigenvalues  $\nu_i^2$ , which we number so that  $0 < \nu_1^2 \leq \dots \leq \nu_r^2$ . Let  $B^*By_i = \nu_i^2 y_i$ . One finds that the partitioned vectors  $(By_i, \pm \nu_i y_i)$  are  $2r$  linearly independent eigenvectors of  $Q$  belonging to the  $2r$  eigenvalues  $1 \pm \nu_i$  ( $i=1, \dots, r$ ). Since  $Q$  is definite, all  $\nu_i < 1$ .

If  $p-r > 0$ , there are  $p-r$  linearly independent vectors  $u_j$  with  $B^*u_j = 0$ . Then  $(u_j, 0)$  are  $p-r$  linearly independent eigenvectors of  $Q$  belonging to the eigenvalue 1. Similarly, if  $q-r > 0$ , there are  $q-r$  linearly independent eigenvectors of  $Q$  of type  $(0, v_k)$ , which all belong to the eigenvalue 1. Here all  $Bv_k = 0$ .

We have found all  $p+q$  eigenvalues of  $Q$ , and see that the largest is  $\Lambda = 1 + \nu_r$ , with an eigenvector  $(By_r, \nu_r y_r)$ . The smallest is  $\lambda = 1 - \nu_r$ , with an eigenvector  $(By_r, -\nu_r y_r)$ . Theorem 3 then completes the proof.

For any scalar  $c$ ,  $P(cD^*QD) = P(D^*QD)$ . It would be interesting to know, for the  $Q$  of (12), when the class  $cQ$  contains all the best conditioned transforms  $D^*QD$ . These transforms essentially constitute the matrices with Young's Property A [3], often encountered in the numerical solution of partial differential equations.

We can show that the partitioned positive definite matrices

$$\begin{bmatrix} c_1 I_p & B \\ B^* & c_2 I_q \end{bmatrix}$$

are best conditioned if and only if  $c_1 = c_2$ . On the other hand, the third order matrices

$$Q = \begin{bmatrix} 1 & 0 & b \\ 0 & d & 0 \\ b & 0 & 1 \end{bmatrix},$$

where  $|b| < 1$  and  $1 - |b| \leq d \leq 1 + |b|$ , are all best conditioned, with  $P(Q) = (1 + |b|) \cdot (1 - |b|)^{-1}$ . We conjecture that, for  $Q$  as in (12), any best conditioned matrix  $D^*QD \neq cQ$  has the form

$$\Pi^* \begin{bmatrix} cI_{p_1} & 0 & B_1 \\ 0 & D_{p_2} & 0 \\ B_1^* & 0 & cI_{q_2} \end{bmatrix} \Pi$$

where  $D_{p_2}$  is diagonal and  $\Pi$  is a permutation matrix.

#### REFERENCES

1. John Todd, *The condition of a certain matrix*, Proc. Cambridge Philos. Soc. vol. 46 (1950) pp. 116-118.
2. David Young, *On the solution of linear systems by iteration*, Proceedings of the Sixth Symposium in Applied Mathematics, to be published.
3. ———, *Iterative methods for solving partial difference equations of elliptic type*, Trans. Amer. Math. Soc. vol. 76 (1954) pp. 92-111.

UNIVERSITY OF CALIFORNIA, LOS ANGELES