

## THE NUMERICAL APPROXIMATION OF THE EIGENVALUES AND EIGENVECTORS FOR SYMETRICAL MATRICES

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### INTRODUCTION

One of the oldest ideas for the calculation of the eigenvalues and eigenvectors of the dense, symmetric matrices is Jacobi's algorithm introduced in 1845.

The procedure (computation method) focused specialists' attention as the parallel calculation techniques were developing.

The paper presents Jacobi's rotation method and the corresponding program within the programming MATLAB medium for approximate calculation of eigenvalues and eigenvectors of the symmetric matrices.

### 1. JACOBI'S ROTATION METHOD

Let  $A$  be a square symmetric matrix of order  $n$ ; then its eigenvalues are real and there exists an orthonormal basis made of the eigenvectors  $v_i$ ,  $i = \overline{1, n}$  in which the equations  $Av_i = \lambda_i v_i$ ,  $i = \overline{1, n}$ , hold and the matrix gets the diagonal form

$$D = \begin{pmatrix} \lambda_1 & 0 & 0 & \dots & 0 \\ 0 & \lambda_2 & 0 & \dots & 0 \\ \dots & \dots & \dots & \dots & \dots \\ 0 & 0 & 0 & \dots & \lambda_n \end{pmatrix} \quad (1)$$

The basis can be chosen so as to lead to  $\lambda_1 \geq \lambda_2 \geq \lambda_3 \dots \geq \lambda_n$ .

If, moreover, matrix  $A$  is definitely positive, then  $\lambda_1 \geq \lambda_2 \geq \lambda_3 \geq \dots \geq \lambda_n > 0$  and

$$\lambda_1 = \|A\|_2 = \sup_{x \neq 0} \frac{\langle Ax, x \rangle}{\langle x, x \rangle}.$$

Lets  $P$  be the transformation matrix from the standard basis of the space  $R^n$  to the basis  $(v_1, v_2, \dots, v_n)$ . Once can easily see that  $P^T P = I$ ; hence  $P$  is orthogonal. It follows that  $P^{-1} = P^T$  and that  $D = P^T A P$ .

In practice, the eigenvalues of matrix  $A$  cannot be actually identified by numerically solving the characteristic equation  $\det(A - \lambda I) = 0$ , because the roots of this polynomial are very "sensitive" to any modification in the coefficients of the polynomial.

The recommended method is to bring, by any possible means, the matrix to a diagonal form and then the eigenvalues can be wholly identified, since they are the elements on the main diagonal of  $D$ .

Thus we aim, through similarity changes that do not modify the eigenvalues, to diminish, possibly up to the total vanishing, the non-diagonal elements of the matrix; thus we could eventually obtain the diagonal matrix.

Jacobi's method consists in performing a series of similarity transformations on the  $A$  matrix by using the simplest nontrivial orthogonal matrices (rotation matrices) of the form

$$U = \begin{pmatrix} p & q & & & \\ & q & p & & \\ & & & \ddots & \\ & & & & \ddots \\ & & & & & \cos \varphi & \dots & \sin \varphi & \dots & 0 \\ & & & & & \cdot & 1 & \cdot & & \\ & & & & & \vdots & \vdots & \ddots & \vdots & \vdots \\ & & & & & & & -\sin \varphi & \dots & \cos \varphi & \dots & 0 \\ & & & & & & & \cdot & \cdot & \cdot & 1 & 0 \\ & & & & & & & & & & 0 & \dots & \dots & 0 & 1 \end{pmatrix}. \quad (2)$$

The elements of matrix  $U$  are given by

$$\begin{cases} u_{ii} = 1 \text{ if } i \neq p \text{ and } i \neq q \\ u_{pp} = \cos \varphi, u_{pq} = \sin \varphi \\ u_{qp} = -\sin \varphi, u_{qq} = \cos \varphi \\ u_{ij} = 0 \text{ as remainder} \end{cases} \quad (3)$$

The matrix  $U$  is orthogonal ( $U^T U = I$  so  $U^{-1} = U^T$ ) and, from a geometrical viewpoint, it represents a rotation of angle  $\varphi$  in the plane identified by the  $e_p$  and  $e_q$  directions.

We denote  $A' = U^T A$  and  $A'' = A' U = U^T A U$ . In general, the elements of matrix  $A'$  are

$$\begin{cases} a'_{ij} = a_{ij} \text{ if } i \neq p \text{ and } j \neq q \\ a'_{pj} = a_{pj} \cdot \cos \varphi - a_{qj} \cdot \sin \varphi \\ a'_{qj} = a_{pj} \cdot \sin \varphi + a_{qj} \cdot \cos \varphi \end{cases} \quad (4)$$

and the ones of matrix  $A''$  are

$$\begin{cases} a''_{ij} = a'_{ij} \text{ if } i \neq p \text{ and } j \neq q \\ a''_{ip} = a'_{ip} \cdot \cos \varphi - a'_{iq} \cdot \sin \varphi \\ a''_{iq} = a'_{ip} \cdot \sin \varphi + a'_{iq} \cdot \cos \varphi \end{cases} \quad (5)$$

It follows from (4) and (5) that

$$\begin{cases} a''_{pp} = a_{pp} \cos^2 \varphi - 2a_{pq} \cos \varphi \sin \varphi + a_{qq} \sin^2 \varphi \\ a''_{qq} = a_{pp} \sin^2 \varphi + 2a_{pq} \cos \varphi \sin \varphi + a_{qq} \cos^2 \varphi \\ a''_{pq} = (a_{pp} - a_{qq}) \sin \varphi \cos \varphi + a_{pq} \cos 2\varphi \\ a''_{qp} = a''_{pq} \end{cases} \quad (6)$$

We aim that largest (in absolute value) non-diagonal element, the should vanish, as a result of the rotation; we will choose the rows  $p$  and, so that  $a_{pq}$  will be the largest element (in absolute value) above the main diagonal and we shall state the condition that  $a''_{pq} = 0$ .

Taking into account (6) it follows that

$$\frac{1}{2} (a_{pq} - a_{qq}) \sin 2\varphi + a_{pq} \cos 2\varphi = 0$$

and therefore

$$\operatorname{tg} 2\varphi = \frac{2a_{pq}}{a_{qq} - a_{pp}} \quad (7)$$

Hence, the angle of rotation is identified from (7). We will introduce the notations:

$$\theta = \frac{a_{qq} - a_{pp}}{2a_{pq}} \text{ and } \operatorname{tg} \varphi = t \quad (8)$$

Since  $\operatorname{tg} 2\varphi = \frac{2\operatorname{tg} \varphi}{1 - \operatorname{tg}^2 \varphi}$ , it results from (7) and (8) that  $t^2 + 2\theta t - 1 = 0$ .

By solving this equation we get

$$t_{1,2} = -\theta \pm \sqrt{\theta^2 + 1} = \frac{1}{\theta \pm \sqrt{\theta^2 + 1}}$$

In order to avoid the case when, the denominator becomes very small we consider

$$t = \begin{cases} \frac{1}{\theta + \operatorname{sgn}(\theta) \sqrt{\theta^2 + 1}} \text{ if } \theta \neq 0 \\ 1 \text{ if } \theta = 0 \end{cases} \quad (9)$$

According to some elementary trigonometric formulas, we have:

$$\begin{cases} c = \cos \varphi = \frac{1}{\sqrt{1+t^2}} \\ s = \sin \varphi = \frac{t}{\sqrt{1+t^2}} \end{cases} \quad (10)$$

From (9) and (10) it follows that  $|t| \leq 1$ ,

$$c \geq \frac{1}{\sqrt{2}}, \quad |s| \leq \frac{1}{\sqrt{2}} \text{ and thus } \varphi \in \left[ -\frac{\pi}{4}, \frac{\pi}{4} \right].$$

If we denote  $S(C)$  the sum of the squares of the non-diagonal elements of any matrix  $C$  then, from (4) and (5), a direct calculation would lead us to:

$$S(A'') = S(A) - 2a_{pq}^2 + 2a_{pq}''^2.$$

Therefore, if we choose the angle of rotation  $\varphi$  according to (9) and (10), it results that

$$a_{pq}'' = 0 \text{ and thus}$$

$$S(A'') = S(A) - 2a_{pq}^2 \quad (11)$$

Since  $a_{ij}^2 \leq a_{pq}^2$  for  $i \neq j$ , we get

$$S(A) \leq n(n-1)a_{pq}^2 \text{ or} \quad (12)$$

$$\frac{-2}{n(n-1)}S(A) \geq -2a_{pq}^2$$

We thus get, from (11) and (12),

$$S(A'') \leq S(A)\left(1 - \frac{2}{n(n-1)}\right) \text{ for } n \geq 2 \quad (13)$$

Let us now consider a long series of rotations leading to the matrices  $A_0, A_1, A_2, \dots, A_k$  where  $A_0 = A, A_1 = A'', A_2 = A_1''$ , etc

It follows from (13) that

$$S(A_k) \leq \left(1 - \frac{2}{n(n-1)}\right)^k S(A) \quad (14)$$

But  $1 - \frac{2}{n(n-1)} \in (0, 1)$  for  $n > 2$ , hence

according to (14),

$$\lim_{k \rightarrow \infty} S(A_k) = 0$$

Thus, at the limit, the sequence  $\{A_k\}$  tends to the diagonal matrix. We can prove the theorem that follows. The matrix  $P$  is the result of product matrices  $U_1, U_2, \dots, U_k$ .

### **Theorem** [3]

Let us consider the eigenvalues  $\lambda_j$  of matrix  $A$  and let  $a_{ij}^{(k)}$  be the diagonal elements of the matrix  $A_k$ . Then:

$$|a_{ij}^{(k)} - \lambda_j| \leq \sqrt{S(A_k)}.$$

Since  $a_{pq}^{(k)}$  is the largest (in absolute value) non-diagonal element of the  $A_k$  matrix, the following evaluation results:

$$S(A_k) \leq (n^2 - n) (a_{pq}^{(k)})^2 < n^2 (a_{pq}^{(k)})^2.$$

From this theorem we obtain :

$$|a_{ij}^{(k)} - \lambda_j| < n |a_{pq}^{(k)}|. \quad (15)$$

The inequality (15) can be considered as a criterion for stopping the algorithm.

From inequality  $n |a_{pq}^{(k)}| < \text{eps}$ , we will get the number  $k$  of the necessary rotations, to approximate the eigenvalues  $\lambda_j$  of the matrix  $A$ , with the diagonal elements  $a_{ii}^{(k)}$  of matrix  $A_k$ .

The sequence of matrices  $A_k$  and  $P$  the transformation matrix, are recursively calculated by

$$\begin{cases} A_k = U_k^T A_{k-1} U_k, & k \geq 1 \\ A_0 = A \end{cases} \quad (16)$$

$$P := U_k P, \quad k \geq 1 \quad (17)$$

The columns of the matrix  $P$  are the eigenvectors of matrix  $A$ .

## 2. THE COMPUTER PROGRAM

Algorithm for the identification of eigenvalues through Jacobi's rotation procedure

Introduce  $A$ ,  $\text{eps}$ ;

Repeat

Determine: **max**: = the highest element in absolute value, above the main diagonal of matrix  $A$ ;  
( $p, q$ ): the position of this element;

Calculate  $A := U^T * A * U$ ;

Calculate  $P := U * P$ ;

Calculate  $s := \left( \sum_{i=1}^n \sum_{\substack{j=1 \\ i \neq j}}^n a_{ij}^2 \right)^{\frac{1}{2}}$  until  $s < \text{eps}$ .

% Jacobi method for calculate eigenvalues and eigenvectors.

% We introduce  $A, s, n, \text{eps}, \text{max1}$ .

$m = 0$ ;

$P = \text{eye}(\text{size}(A))$ ;

while  $s > \text{eps}$

for  $r = 1 : n-1$

for  $k = r+1 : n$

if  $\text{max1} < \text{abs}(A(r,k))$

$\text{max1} = \text{abs}(A(r,k))$ ;

$p = r$ ;

$q = k$ ;

end

end

end

if  $A(p, q) == 0$

disp('STOP')

break

end

$teta = (A(q,q) - A(p,p))/2 * A(p,q)$ ;

if  $teta == 0$

$t = 1$ ;

else

$t = 1 / (teta + \text{sign}(teta) * (teta^2 + 1)^{(1/2)})$ ;

end

$c = 1 / (t^2 + 1)^{(1/2)}$ ;

$s = t / (t^2 + 1)^{(1/2)}$ ;

$I = \text{eye}(\text{size}(A))$ ;

$I(p,p) = c$ ;

$I(q,q) = c$ ;

$I(p,q) = s$ ;

$I(q,p) = -s$ ;

disp('the matrix U')

disp(I)

$P = P * I$ ;

disp('the matrix P')

disp(P)

$A = I' * A * I$ ;

disp('the matrix A')

disp(A)

$s = 0$ ;

for  $r = 1 : n$

for  $k = 1 : n$

if  $r \neq k$

$s = s + A(r,k)^2$ ;

end

end

end

$s = s^{(1/2)}$

$m = m + 1$ ;

disp('number iteration')

disp(m)

$\text{max1} = 0$ ;

end

% The significance of the variables

% A symmetric square matrix

%  $n$  – the order of matrix A

%  $\text{max1}$  – contains the highest  $a(p,q)$  in absolute

% value

%  $\text{eps}$  – increased for  $s$

%  $P$  – the transformation matrix ; at the end, its

% columns contain the eigenvectors of matrix A

% At the end, on the main diagonal of the matrix,

% one can read the eigenvalues of matrix A.

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