

## Task 9. Calculation of the matrix eigenvector for the given eigenvalue

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## Task 9

Write a program that calculates eigenvector of the matrix  $\mathbf{A}$  for the given eigenvalue.

## Eigenvalue problem

- ▶ Eigenvalue equation for the square matrix  $n \times n$

$$\mathbf{A}\mathbf{x}_k = \lambda_k\mathbf{x}_k \quad \Leftrightarrow \quad (\mathbf{A} - \lambda_k\mathbf{1}_n)\mathbf{x}_k = 0 \quad k = 1, 2, \dots, n$$

where  $\mathbf{x}_i$  is the eigenvector ( $n \times 1$ ) of  $\mathbf{A}$  matrix belonging to  $\lambda_i$  eigenvalue being the number, in general, the complex number;  $\mathbf{1}_n$  is the  $n \times n$  unit matrix

- ▶ Definition:  $\mathbf{X} = [\mathbf{x}_1 \dots \mathbf{x}_n]$  is the matrix  $n \times n$ , where  $\mathbf{x}_k$  its columns  $n \times 1$  are the eigenvectors of the matrix  $\mathbf{A}$  corresponding to consecutive eigenvalues  $\lambda_1, \dots, \lambda_n$ . It is noted that  $\det \mathbf{X} \neq 0$ , because eigenvectors  $\mathbf{x}_k$  are linearly independent
- ▶ Spectral decomposition of the matrix  $\mathbf{A}$

$$\mathbf{A} = \mathbf{X}\mathbf{D}\mathbf{X}^{-1},$$

where  $\mathbf{D}$  is the diagonal matrix  $(\mathbf{D})_{ij} = \lambda_i\delta_{ij}$  and  $\delta_{ij}$  is the Kronecker delta

- ▶ Spectral decomposition of the inverse matrix  $\mathbf{A}^{-1}$

$$\mathbf{A}^{-1} = \mathbf{X}\mathbf{D}^{-1}\mathbf{X}^{-1},$$

where  $\mathbf{D}^{-1}$  is the diagonal matrix and  $(\mathbf{D}^{-1})_{ij} = \lambda_i^{-1}\delta_{ij}$

## (In the context of) shifted inverse iteration method

- ▶ Eigenvalue equation of the shifted matrix  $\mathbf{A}_s = \mathbf{A} - s\mathbf{1}_n$ , where  $s \in \mathbb{R}$  and  $\mathbf{A}\mathbf{x}_k = \lambda_k\mathbf{x}_k$

$$\mathbf{A}_s\mathbf{x}_k = \mathbf{A}\mathbf{x}_k - s\mathbf{1}_n\mathbf{x}_k = \lambda_k\mathbf{x}_k - s\mathbf{x}_k = (\lambda_k - s)\mathbf{x}_k$$

Shifted matrix  $\mathbf{A}_s$  possesses the same eigenvectors as the matrix  $\mathbf{A}$ , and the eigenvalues of  $\mathbf{A}_s$  are “shifted” by  $s$ , i.e.  $\lambda_s = \lambda_k - s$

- ▶ Spectral decomposition of  $\mathbf{A}_s$  and  $\mathbf{A}_s^{-1}$  matrices

$$\mathbf{A}_s = \mathbf{X}\mathbf{D}_s\mathbf{X}^{-1} \quad (\mathbf{D}_s)_{ij} = (\lambda_i - s)\delta_{ij}$$

$$\mathbf{A}_s^{-1} = \mathbf{X}\mathbf{D}_s^{-1}\mathbf{X}^{-1} \quad (\mathbf{D}_s^{-1})_{ij} = \frac{\delta_{ij}}{\lambda_i - s}$$

- ▶ Consider *sample* vector  $\mathbf{x}^{(0)} = \sum_{k=1}^n c_k\mathbf{x}_k$  <sup>(1)</sup>, then

$$\mathbf{A}_s^{-1}\mathbf{x}_k = \frac{1}{\lambda_k - s}\mathbf{x}_k \quad \Rightarrow \quad \mathbf{A}_s^{-1}\mathbf{x}^{(0)} = \sum_{k=1}^n \frac{c_k}{\lambda_k - s}\mathbf{x}_k$$

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<sup>1</sup>Eigenvectors  $\mathbf{x}_k$  are linearly independent, thus any vector can be expressed as their linear combination.

## (In the context of) shifted inverse iteration method - cont.

- ▶ If matrix  $\mathbf{A}$  possesses different eigenvalues, i.e.  $\lambda_1 > \lambda_2 > \dots > \lambda_n$ , then for  $s \approx \lambda_i$  we have

$$\left| \frac{1}{\lambda_i - s} \right| \gg \left| \frac{1}{\lambda_k - s} \right| \quad k \neq i$$

$$\mathbf{A}_s^{-1} \mathbf{x}^{(0)} = \sum_{k=1}^n \frac{c_k}{\lambda_k - s} \mathbf{x}_k \approx \frac{c_i}{\lambda_i - s} \mathbf{x}_i \equiv \frac{c_i}{\epsilon} \mathbf{x}_i,$$

where  $\lambda_i - s \equiv \epsilon$ ; **it is crucial that  $c_i \neq 0$**

- ▶ If  $|c_i| \sim 1$ ,  $|\epsilon| \ll 1$  and  $\|\mathbf{x}^{(0)}\| \sim 1$ , then for  $s \rightarrow \lambda_i$ , i.e.  $\epsilon \rightarrow 0$ , we have

$$\mathbf{A}_s \mathbf{x}_i = \frac{\epsilon}{c_i} \mathbf{x}^{(0)} \longrightarrow \mathbf{0}_{n \times 1},$$

because this limit leads to  $\mathbf{A}_s \rightarrow \mathbf{A}_{\lambda_i}$  and, as a consequence,

$$\mathbf{A}_{\lambda_i} \mathbf{x}_i = (\mathbf{A} - \lambda_i \mathbf{1}_n) \mathbf{x}_i = \mathbf{0}_{n \times 1}$$

# Algorithm

1. Calculate  $\mathbf{A}_s = \mathbf{A} - s\mathbf{1}_n$ , where  $s = \lambda_i - \epsilon$  and  $\epsilon \sim 0$
2. Perform decomposition  $\mathbf{A}_s = \mathbf{LUP}$  (Doolittle's method)
3. For the sample vector  $\mathbf{x}^{(0)}$ , where  $\|\mathbf{x}^{(0)}\| = 1$  solve the equation

$$\mathbf{A}_s \mathbf{y}^{(1)} = \mathbf{x}^{(0)}$$

(apply LUP decomposition)

4. Normalize  $\mathbf{y}^{(1)}$ , i.e.  $\mathbf{x}^{(1)} = \frac{\mathbf{y}^{(1)}}{\|\mathbf{y}^{(1)}\|}$ , then  $\|\mathbf{x}^{(1)}\| = 1$ ;  $\mathbf{x}^{(1)}$  is the eigenvector  $\mathbf{x}_i$  (or its subsequent approximation)
5. In general, points 3 and 4 should be repeated

## Remarks

- ▶ Matrix  $\mathbf{A}_s$  is almost singular for  $\epsilon \sim 0$
- ▶  $\det \mathbf{A}_s \sim 0$ , thus in Doolittle's method  $\det \mathbf{L} = 1$  and  $\det \mathbf{U} \sim 0$
- ▶ Minimal value of  $\epsilon \sim \lambda_i \epsilon_{mach} \neq 0$ , because we want to avoid  $\det \mathbf{U} = 0$ ; on the other hand we want to have single-step method, which is possible for  $\epsilon \sim 0$
- ▶  $\|\mathbf{y}\| \sim \frac{1}{|\epsilon|} \gg 1$ , e.g. in double precision  $\|\mathbf{y}\| \sim 10^{15}$ , then
$$\|\mathbf{A}_s \mathbf{x}^{(1)}\| = \frac{\|\mathbf{x}^{(0)}\|}{\|\mathbf{y}^{(1)}\|} \sim 10^{-15}$$
- ▶ Sample input vector has to be such that  $\mathbf{x}^{(0)T} \mathbf{x}_i \neq 0$

## Other approach

- ▶ Assume that  $\mathbf{A}$  possesses different eigenvalues
- ▶  $\mathbf{A}_{\lambda_i} \mathbf{x}_i = \mathbf{0}_{n \times 1}$
- ▶  $\mathbf{A}_{\lambda_i}$  is singular, i.e.  $\det \mathbf{A}_{\lambda_i} = 0$
- ▶ Doolittle's LUP decomposition with partial pivoting

$$\mathbf{A}_{\lambda_i} = \mathbf{LUP},$$

where  $\det \mathbf{L} = 1$ ,  $\det \mathbf{U} = 0$ , the last row in  $\mathbf{U}$  is filled with zeros

- ▶ System of linear equations for  $\mathbf{x}_i$ , i.e.  $\mathbf{A}_{\lambda_i} \mathbf{x}_i = \mathbf{0}_{n \times 1}$ , possesses infinite number of solutions and (at least) one arbitrary parameter
- ▶ Denote  $\mathbf{P}\mathbf{x}_i = \mathbf{x}'_i$ ,  $\mathbf{U}\mathbf{x}'_i = \mathbf{y}_i$ , then

$$\mathbf{A}_{\lambda_i} \mathbf{x}_i = \mathbf{L}\mathbf{y}_i = \mathbf{0}_{n \times 1} \Rightarrow \mathbf{U}\mathbf{x}'_i = \mathbf{0}_{n \times 1}$$

(this is because  $\mathbf{L}$  is not singular and, as a consequence,  $\mathbf{y}_i = \mathbf{0}_{n \times 1}$ )

## Algorithm "2"

1. Calculate  $\mathbf{A}_{\lambda_i} = \mathbf{A} - \lambda_i \mathbf{1}_n$
2. Decompose  $\mathbf{A}_{\lambda_i} = \mathbf{LUP}$  by means of Doolittle's method with partial pivoting
3. Solve the system of linear equations

$$\mathbf{U}\mathbf{x}'_i = \mathbf{0}_{n \times 1}$$

by setting  $(\mathbf{x}'_i)_n = 1$  the remaining elements  $(\mathbf{x}'_i)_k$ ,  $k = 1, \dots, n - 1$  may be calculated with backward substitution method

4. Eigenvector  $\mathbf{x}_i = \mathbf{P}^{Tr} \mathbf{x}'_i$
5. We can normalize the eigenvector, i.e.  $\mathbf{x}_i \rightarrow \frac{\mathbf{x}_i}{\|\mathbf{x}_i\|}$

# TO DO

1. Read the real matrix  $\mathbf{A}_{n \times n}$  and its eigenvalues
2. Calculate eigenvectors  $\mathbf{x}_i$  of  $\mathbf{A}$  matrix for the corresponding eigenvalues  $\lambda_i$
3. Check  $\mathbf{A}\mathbf{x}_i = \lambda_i\mathbf{x}_i$
4. Calculate eigenvectors for symmetric and non-symmetric matrices
5. Check the performance of the method for the matrix that possesses to the same eigenvalues

## Examples

$$A = \begin{pmatrix} 1 & 1 & 1 \\ 2 & -2 & 2 \\ 3 & 3 & -3 \end{pmatrix}$$

$$\lambda = 2.76300328661375, -1.72368589498208, -5.03931739163167$$

$$A = \begin{pmatrix} 3 & 0 & 2 & -2 \\ 2 & 0 & -2 & 2 \\ 0.475 & -0.65 & 4.5 & -1.625 \\ 1.1 & -1.4 & 0 & 2.5 \end{pmatrix}$$

$$\lambda = 1, 2, 3, 4$$

$$A = \begin{pmatrix} 1 & 2 & 3 & 5 \\ 2 & 3 & 4 & 5 \\ 3 & 4 & 5 & 6 \\ 5 & 5 & 6 & 8 \end{pmatrix}$$

$$\lambda = -1.7292612617663754, -0.0437773119849116, 0.7322067668156992, \\ 18.0408318069355893$$

$$A = \begin{pmatrix} 2 & 13 & -14 & 3 \\ -2 & 25 & -22 & 4 \\ -3 & 31 & -27 & 5 \\ -2 & 34 & -32 & 7 \end{pmatrix}$$

$$\lambda = 1, 1, 2, 3$$