

# DIAGONAL MATRICES

## THE EIGNEVALUES OF A MATRIX

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# OUTLINE

Notion of Eigenvalue and Eigenvector

Definition

Determine the eigenvalues and the eigenvectors

Basis transformation

Characteristic polynomial

Diagonal matrix

Symmetric matrices

Back to the quadratic forms

Problems

# IMPORTANT

Taking part on the written exam at 4.00 pm. of December 16, 2025 is compulsory!

# NOTION OF EIGENVALUE AND EIGENVECTOR

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# NOTION OF EIGENVALUE AND EIGENVECTOR

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## DEFINITION

# EIGENVALUE AND EIGENVECTOR

## Definition

The scalar  $\lambda \in \mathbb{R}$  is called an **eigenvalue** of the matrix  $\mathbf{A} \in \mathbb{R}^{n \times n}$ , if there exists a nonzero vector  $\mathbf{s} \in \mathbb{R}^n$  for which

$$\mathbf{A}\mathbf{s} = \lambda\mathbf{s}$$

holds. The vector  $\mathbf{s}$  is said to be the **eigenvector** corresponding to the eigenvalue  $\lambda$  of matrix  $\mathbf{A}$ .

## THE SET OF EIGENVECTORS FORM A SUBSPACE

If both  $\mathbf{s}$  and  $\mathbf{t}$  are eigenvectors corresponding to the eigenvalue  $\lambda$  then their any nontrivial linear combination  $\alpha\mathbf{s} + \beta\mathbf{t}$  is also an eigenvector with the same eigenvalue  $\lambda$ :

$$\mathbf{A}(\alpha\mathbf{s} + \beta\mathbf{t}) = \alpha\mathbf{A}\mathbf{s} + \beta\mathbf{A}\mathbf{t} = \alpha\lambda\mathbf{s} + \beta\lambda\mathbf{t} = \lambda(\alpha\mathbf{s} + \beta\mathbf{t}).$$

## DEPENDENT SYSTEM

The eigenvectors corresponding to the eigenvalue  $\lambda$  are the nonzero solutions of the homogeneous system of linear equations

$$(\mathbf{A} - \lambda \mathbf{E}) \mathbf{x} = \mathbf{0}.$$

A homogeneous system of linear equations with a square coefficient matrix has nonzero solution if and only if its columns form a dependent system.

It follows that  $\lambda$  is an eigenvalue of  $\mathbf{A}$  if and only if the columns of matrix

$$\mathbf{A} - \lambda \mathbf{E}$$

is a linearly dependent system.



## DETERMINE THE EIGENVALUES AND THE EIGENVECTORS

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### Example

Find the eigenvalues and the corresponding eigenvectors of the following matrix:

$$\mathbf{A} = \begin{bmatrix} 1 & 0 & -1 \\ 0 & 1 & 1 \\ 2 & 0 & -2 \end{bmatrix}$$

We find the values of the parameter  $t$  for which the matrix  $\mathbf{A} - t\mathbf{E}$  is singular and with the obtained values we find the nonzero solutions of the homogeneous system of linear equations  $(\mathbf{A} - t\mathbf{E})\mathbf{x} = \mathbf{0}$ .

# DETERMINE THE EIGENVALUES AND THE EIGENVECTORS

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## BASIS TRANSFORMATION

# FIND THE EIGENVALUES AND THE EIGENVECTOR WITH BASIS TRANSFORMATION

Using elementary basis transformation we solve the homogeneous system of linear equations  $(\mathbf{A} - t\mathbf{E})\mathbf{x} = \mathbf{0}$  for each value of the parameter  $t$  for which the coefficient matrix is singular:

$$\begin{array}{c|ccc} & x_1 & x_2 & x_3 \\ \hline & 1-t & 0 & -1 \\ & 0 & 1-t & 1 \\ & 2 & 0 & -2-t \end{array} \rightarrow \begin{array}{c|cc} & x_1 & x_2 \\ \hline x_3 & t-1 & 0 \\ & 1-t & 1-t \\ & t^2+t & 0 \end{array}$$

If  $t = 1$  then we can continue as follows:

$$\begin{array}{c|cc} & x_1 & x_2 \\ \hline x_3 & 0 & 0 \\ & 0 & 0 \\ & \boxed{2} & 0 \end{array} \rightarrow \begin{array}{c|c} & x_2 \\ \hline x_3 & 0 \\ & 0 \\ x_1 & 0 \end{array}$$

From this table we obtain that at  $t = 1$  the coefficient matrix is singular and the solution of the homogeneous system of linear equations  $(\mathbf{A} - \mathbf{E})\mathbf{x} = \mathbf{0}$

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \tau \quad \tau \in \mathbb{R},$$

that is 1 is an eigenvalue and the above vectors are the corresponding eigenvectors for all  $\tau \neq 0$ .

If  $t \neq 1$  then

$$\begin{array}{c|cc} & x_1 & x_2 \\ \hline x_3 & t-1 & 0 \\ & 1-t & \boxed{1-t} \\ & t^2+t & 0 \end{array} \rightarrow \begin{array}{c|c} & x_1 \\ \hline x_3 & t-1 \\ x_2 & 1 \\ & t^2+t \end{array}$$

and we obtain that the coefficient matrix is also singular if  $t = 0$  or  $t = -1$ .

One can see, from the table, that the eigenvectors corresponding to the eigenvalue 0 are

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix} \tau \quad \tau \in \mathbb{R} \quad \tau \neq 0$$

and the eigenvectors corresponding to the eigenvalue  $-1$  are

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 1 \\ -1 \\ 2 \end{bmatrix} \tau \quad \tau \in \mathbb{R} \quad \tau \neq 0.$$

There are real matrices which has no real eigenvalue, for instance the matrix

## Matrix having no real eigenvalue

$$\mathbf{A} = \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix}$$

has no real eigenvalue, because the columns of matrix  $A - tE$  always form an independent system.

# CHARACTERISTIC POLYNOMIAL

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## Definition

If  $A$  is a square matrix then the polynomial

$$k(t) = \det(A - tI)$$

is called the *characteristic polynomial* of the matrix  $A$ .

## Theorem

*Let  $A$  be a square matrix and consider a real number  $\lambda$ . The number  $\lambda$  is an eigenvalue of matrix  $A$  if and only if  $k(\lambda) = 0$ .*

Verify that similar matrices have the same characteristic polynomial.

## TWO ALTERNATIVES LOOKING FOR EIGENVALUES AND EIGENVECTORS

Now there are two ways to find the eigenvalues and the corresponding eigenvectors:

First We find the values of the parameter  $t$  for which the matrix  $\mathbf{A} - t\mathbf{E}$  is singular and with the obtained values we find the nonzero solutions of the homogeneous system of linear equations  $(\mathbf{A} - t\mathbf{E})\mathbf{x} = \mathbf{0}$ , or

## TWO ALTERNATIVES LOOKING FOR EIGENVALUES AND EIGENVECTORS

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**First** We find the values of the parameter  $t$  for which the matrix  $\mathbf{A} - t\mathbf{E}$  is singular and with the obtained values we find the nonzero solutions of the homogeneous system of linear equations  $(\mathbf{A} - t\mathbf{E})\mathbf{x} = \mathbf{0}$ , or

**Second** We determine the roots of the characteristic polynomial of  $\mathbf{A}$ , those are the eigenvalues of  $\mathbf{A}$  and then solving the homogeneous system of linear equations  $(\mathbf{A} - \lambda\mathbf{E})\mathbf{x} = \mathbf{0}$  for each root we obtain the corresponding eigenvectors.

## 2ND METHOD TO FIND THE EIGENVALUES OF A MATRIX

The characteristic equation:

$$|\mathbf{A} - t\mathbf{E}| = \begin{vmatrix} 1-t & 0 & -1 \\ 0 & 1-t & 1 \\ 2 & 0 & -2-t \end{vmatrix} = -t^3 + t = 0$$

Its solutions:  $t_1 = 0$ ,  $t_2 = -1$  and  $t_3 = 1$  are the eigenvalues of the matrix  $\mathbf{A}$ .

The eigenvectors belonging to the eigenvalue 0 are the nonzero solutions of the homogeneous system of linear equations  $(A - 0 \cdot \mathbf{E}) \mathbf{x} = \mathbf{0}$ . Thus the system to be solved:

$$\begin{bmatrix} 1 & 0 & -1 \\ 0 & 1 & 1 \\ 2 & 0 & -2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}.$$

The solutions:  $x_1 = \tau, x_2 = -\tau, x_3 = \tau$  ( $\tau \in \mathbb{R}$ ). Therefore the eigenvectors belonging to the eigenvalue  $t_1 = 0$  are

$$\begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix} \tau, \quad (\tau \in \mathbb{R} \setminus \{0\})$$

The eigenvectors belonging to the eigenvalue  $-1$  are the nonzero solutions of the homogeneous system of linear equations  $(A + 1 \cdot \mathbf{E}) \mathbf{x} = \mathbf{0}$ . Thus the system to be solved:

$$\begin{bmatrix} 2 & 0 & -1 \\ 0 & 2 & 1 \\ 2 & 0 & -1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}.$$

The solutions:  $x_1 = \tau, x_2 = -\tau, x_3 = 2\tau$  ( $\tau \in \mathbb{R}$ ). Therefore the eigenvectors belonging to the eigenvalue  $t_2 = -1$ :

$$\begin{bmatrix} 1 \\ -1 \\ 2 \end{bmatrix} \tau, \quad (\tau \in \mathbb{R} \setminus \{0\}).$$

The eigenvectors belonging to the eigenvalue 1 are the nonzero solutions of the homogeneous system of linear equations  $(A - 1 \cdot \mathbf{E}) \mathbf{x} = \mathbf{0}$ , that is,

$$\begin{bmatrix} 0 & 0 & -1 \\ 0 & 0 & 1 \\ 2 & 0 & -3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}.$$

Solving the system the eigenvectors :

$$\begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \tau, \quad (\tau \in \mathbb{R} \setminus \{0\}).$$

There are real matrices which have no real eigenvalue, for instance the matrix

**Matrix having no real eigenvalue**

$$\mathbf{A} = \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix}$$

has the characteristic polynomial  $t^2 + 1$ , therefore it has no real root.



# DIAGONAL MATRIX

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# DIAGONAL MATRIX

## Definition

A diagonal matrix is a square matrix, whose entries outside of the main diagonal are zeros. They will be denoted by

$$\mathbf{D} = \text{diag} \langle d_1, \dots, d_n \rangle ,$$

$d_1, \dots, d_n$  are the entries in the main diagonal.

The eigenvalues of a diagonal matrix are just the entries in the main diagonal.

# DIAGONALIZABLE MATRIX

## Definition

A square matrix  $\mathbf{A}$  of order  $n$  is said to be diagonalizable if there exists a regular matrix  $\mathbf{P}$  and a diagonal matrix  $\mathbf{D}$  such that

$$\mathbf{P}^{-1}\mathbf{A}\mathbf{P} = \mathbf{D} = \text{diag} \langle d_1, \dots, d_n \rangle.$$

For example the matrix

$$\mathbf{A} = \begin{pmatrix} 3 & 1 \\ 2 & 4 \end{pmatrix}$$

is diagonalizable, because

$$\mathbf{P}^{-1}\mathbf{A}\mathbf{P} = -\frac{1}{3} \begin{bmatrix} 2 & -1 \\ -1 & -1 \end{bmatrix} \begin{bmatrix} 3 & 1 \\ 2 & 4 \end{bmatrix} \begin{bmatrix} -1 & 1 \\ 1 & 2 \end{bmatrix} = \begin{bmatrix} 2 & 0 \\ 0 & 5 \end{bmatrix}$$

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It is fascinating. :) How can we find the matrix  $\mathbf{P}$ , above?

# A NECESSARY AND SUFFICIENT CONDITION

## Theorem

*A square matrix  $\mathbf{A}$  of order  $n$  is diagonalizable if and only if it has  $n$  linearly independent eigenvectors  $\mathbf{x}_1, \dots, \mathbf{x}_n$ . In this case*

$$\mathbf{P}^{-1}\mathbf{A}\mathbf{P} = \begin{bmatrix} \lambda_1 & 0 & \dots & 0 \\ 0 & \lambda_2 & \dots & 0 \\ \vdots & \vdots & & \vdots \\ 0 & 0 & \dots & \lambda_n \end{bmatrix},$$

*where the columns of  $\mathbf{P}$  are  $\mathbf{x}_1, \dots, \mathbf{x}_n$ , and  $\lambda_1, \dots, \lambda_n$  scalars are the corresponding eigenvalues.*

## PROOF

If  $\mathbf{x}_1, \dots, \mathbf{x}_n$  are linearly independent eigenvectors of  $\mathbf{A}$  and  $\lambda_1, \dots, \lambda_n$  are the corresponding eigenvalues, that is,

$$\mathbf{A}\mathbf{x}_1 = \lambda_1\mathbf{x}_1, \quad \mathbf{A}\mathbf{x}_2 = \lambda_2\mathbf{x}_2, \quad \dots, \quad \mathbf{A}\mathbf{x}_n = \lambda_n\mathbf{x}_n.$$

then let  $\mathbf{P} = [\mathbf{x}_1, \dots, \mathbf{x}_n]$ . Thus

$$\begin{aligned} \mathbf{AP} &= \mathbf{A} [\mathbf{x}_1, \dots, \mathbf{x}_n] = [\mathbf{Ax}_1, \dots, \mathbf{Ax}_n] = [\lambda_1\mathbf{x}_1, \dots, \lambda_n\mathbf{x}_n] = \\ &[\mathbf{x}_1, \dots, \mathbf{x}_n] \text{diag} \langle \lambda_1, \dots, \lambda_n \rangle = \mathbf{P} \text{diag} \langle \lambda_1, \dots, \lambda_n \rangle, \end{aligned}$$

therefore multiplying both sides by the inverse of  $\mathbf{P}$ , we obtain the identity required:

$$\mathbf{P}^{-1}\mathbf{AP} = \text{diag} \langle \lambda_1, \dots, \lambda_n \rangle.$$

Conversely, if

$$\mathbf{P}^{-1}\mathbf{A}\mathbf{P} = \text{diag} \langle \lambda_1, \dots, \lambda_n \rangle,$$

then

$$\mathbf{A}\mathbf{P} = \mathbf{P}\text{diag} \langle \lambda_1, \dots, \lambda_n \rangle,$$

that is, for each column  $\mathbf{x}_i$  ( $i = 1, \dots, n$ ) of  $\mathbf{P}$  the equation  $\mathbf{A}\mathbf{x}_i = \lambda_i\mathbf{x}_i$  holds, verifying that  $\lambda_i$  is an eigenvalue of  $\mathbf{A}$  and a corresponding eigenvector is  $\mathbf{x}_i$ .

## Example

Decide if the matrix  $\mathbf{A} = \begin{pmatrix} 3 & 1 \\ 2 & 4 \end{pmatrix}$  is diagonalizable! If it is diagonalizable then give an invertible matrix  $\mathbf{P}$  for which  $\mathbf{P}^{-1}\mathbf{A}\mathbf{P}$  is a diagonal matrix.



## Example

Decide if the matrix  $\mathbf{A} = \begin{pmatrix} 3 & 1 \\ 2 & 4 \end{pmatrix}$  is diagonalizable! If it is diagonalizable then give an invertible matrix  $\mathbf{P}$  for which  $\mathbf{P}^{-1}\mathbf{A}\mathbf{P}$  is a diagonal matrix.

The eigenvalues of  $\mathbf{A}$  are  $\lambda_1 = 2$  and  $\lambda_2 = 5$ , furthermore the corresponding linearly independent eigenvectors:

$$\mathbf{v}_1 = \begin{bmatrix} -1 \\ 1 \end{bmatrix} \quad \text{and} \quad \mathbf{v}_2 = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

## Example

Decide if the matrix  $\mathbf{A} = \begin{pmatrix} 3 & 1 \\ 2 & 4 \end{pmatrix}$  is diagonalizable! If it is diagonalizable then give an invertible matrix  $\mathbf{P}$  for which  $\mathbf{P}^{-1}\mathbf{A}\mathbf{P}$  is a diagonal matrix.

The eigenvalues of  $\mathbf{A}$  are  $\lambda_1 = 2$  and  $\lambda_2 = 5$ , furthermore the corresponding linearly independent eigenvectors:

$$\mathbf{v}_1 = \begin{bmatrix} -1 \\ 1 \end{bmatrix} \quad \text{and} \quad \mathbf{v}_2 = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

Hence by the previous theorem  $\mathbf{A}$  is a diagonalizable matrix.

If  $\mathbf{P} = \begin{bmatrix} -1 & 1 \\ 1 & 2 \end{bmatrix}$ , then

$$\mathbf{P}^{-1}\mathbf{A}\mathbf{P} = -\frac{1}{3} \begin{bmatrix} 2 & -1 \\ -1 & -1 \end{bmatrix} \begin{bmatrix} 3 & 1 \\ 2 & 4 \end{bmatrix} \begin{bmatrix} -1 & 1 \\ 1 & 2 \end{bmatrix} = \begin{bmatrix} 2 & 0 \\ 0 & 5 \end{bmatrix}$$

## AN IMPORTANT SPECIAL CASE

The necessary and sufficient condition for a matrix is being diagonalizable in theorem (7) can not be easily checked. Therefore it is an important result – can be proved using mathematical induction – that eigenvectors of a square matrix  $\mathbf{A}$  corresponding to different eigenvalues are linearly independent, from which it immediately follows that an  $n \times n$  real matrix  $\mathbf{A}$  having  $n$  different real eigenvalues is diagonalizable. Of course the above condition is not a necessary one.

## EXAMPLE

Consider the matrix:

$$\mathbf{A} = \begin{bmatrix} 2 & 1 & 0 \\ 0 & 1 & 0 \\ 0 & 1 & 2 \end{bmatrix}$$

This matrix has only two different eigenvalues  $\lambda_1 = 2$  and  $\lambda_2 = 1$ .

The linearly independent eigenvectors corresponding to the eigenvalue  $\lambda_1 = 2$  are:

$$\mathbf{v}_1 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \quad \text{and} \quad \mathbf{v}_2 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}, \text{ while the eigenvector belonging to the eigenvalue}$$

$$\lambda_2 = 1 \text{ is } \mathbf{v}_3 = \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix}. \text{ Thus the matrix has three linearly independent eigenvectors,}$$

therefore it is diagonalizable.

# SYMMETRIC MATRICES

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# SYMMETRIC MATRICES

The square matrices occurring in economical applications are often symmetric, therefore it seems to be reasonable to study the question if the symmetric matrices are diagonalizable.

Recall that a real matrix  $\mathbf{A}$  is said to be symmetric if  $\mathbf{A}^\top = \mathbf{A}$  and it is orthogonal if  $\mathbf{A}^\top = \mathbf{A}^{-1}$ .

## Theorem

- *The characteristic equation of an  $n \times n$  symmetric matrix has  $n$  real roots counting their multiplicity as well.*
- *The eigenvectors corresponding to different eigenvalues of a symmetric matrix are orthogonal.*

# THE SPECTRAL THEOREM OF SYMMETRIC MATRICES

## Theorem

*The symmetric matrices are diagonalizable, furthermore to each symmetric matrix  $\mathbf{A}$  there exists an orthogonal matrix  $\mathbf{U}$ , such that*

$$\mathbf{U}^{-1}\mathbf{A}\mathbf{U} = \text{diag} \langle \lambda_1, \dots, \lambda_n \rangle ,$$

*where  $\lambda_1, \dots, \lambda_n$  are the eigenvalues of  $\mathbf{A}$ .*

## Example

Find the orthogonal matrix  $\mathbf{U}$  to the symmetric matrix  $\mathbf{A}$  for which the matrix  $\mathbf{U}^\top \mathbf{A} \mathbf{U}$  is a diagonal one.

$$\mathbf{A} = \begin{bmatrix} 3 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix}$$

The eigenvalues of the matrix are  $\lambda_1 = -1$ ,  $\lambda_2 = 1$  and  $\lambda_3 = 3$ , and the corresponding eigenvectors are

$$\mathbf{v}_1 = \begin{bmatrix} 0 \\ -1 \\ 1 \end{bmatrix}, \mathbf{v}_2 = \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix} \quad \text{and} \quad \mathbf{v}_3 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}.$$

The normed eigenvectors are the columns of the orthogonal matrix  $\mathbf{U}$ . Thus

$$\mathbf{U} = \begin{bmatrix} 0 & 0 & 1 \\ -\frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & 0 \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & 0 \end{bmatrix}$$

and  $\mathbf{U}^\top \mathbf{A} \mathbf{U} = \text{diag} \langle -1, 1, 3 \rangle$ .



## BACK TO THE QUADRATIC FORMS

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# CLASSIFICATION WITH EIGENVALUES

## Theorem

*Consider the quadratic form  $Q$ , and let  $B$  denote the corresponding symmetric matrix, i.e.  $Q(x) = \langle x, Bx \rangle$  for every  $x \in \mathbb{R}^n$ . Examine the eigenvalues of  $B$ .*

- If all eigenvalues are positive, then  $Q$  is positive definite.*
- If all eigenvalues are nonnegative and at least one of them is zero, then  $Q$  is positive semidefinite.*
- If all eigenvalues are negative, then  $Q$  is negative definite.*
- If all eigenvalues are nonpositive and at least one of them is zero, then  $Q$  is negative semidefinite.*
- If there are both positive and negative eigenvalues, then  $Q$  is indefinite.*

# CLASSIFICATION WITH LEADING PRINCIPAL MINORS

Let  $B$  denote an  $n \times n$  real matrix. The determinant of the top left  $k \times k$  matrices ( $k = 1, \dots, n$ ) are called **leading principal minors**.

## Theorem (Sylvester)

*Consider the quadratic form  $Q$ , and let  $B$  denote the corresponding symmetric matrix, i.e.*

*$Q(x) = \langle x, Bx \rangle$  for every  $x \in \mathbb{R}^N$ .*

*The quadratic form is positive definite if and only if all of the leading principal minors of  $B$  is positive.*

Assume first, that  $Q$  is positive definite. If  $\lambda$  is an eigenvalue, and  $x$  is an eigenvector such that  $Bx = \lambda x$ , then

$$0 < Q(x) = \langle x, Bx \rangle = \langle x, \lambda x \rangle = \lambda \langle x, x \rangle = \lambda \|x\|^2$$

which means that all of the eigenvalues of  $B$  are positive.

Realize, that the determinant of a symmetric matrix is the product of all of the eigenvalues. We obtained that the determinant of any positive definite matrix is positive.

Obviously, the top left  $k \times k$  submatrix of a positive definite matrix is also positive definite. Thus we proved, that all of the principal minors of  $B$  is a positive number.

Vice versa, assume now that the sequence of the principal minors  $\Delta_1, \dots, \Delta_n$  are all positive numbers. At this case all of the top left  $k \times k$  submatrix is regular, thus all of the columns of any  $k \times k$  submatrix form a linear independent system. This fact implicates, that during the dyadic decomposition process all of the top left pivot terms are nonzero numbers.

If  $\alpha_1, \dots, \alpha_n$  is the sequence of the pivot terms, then the dyadic decompositon is

$$B = \sum_{k=1}^n \alpha_k d_k \cdot d_k^T$$

We know that the derminant can be computed as the product of the pivot terms (at this case) thus  $\Delta_1 = \alpha_1$ , and  $\Delta_k = \alpha_k \Delta_{k-1}$  if  $k > 1$ . This fact implicates that all of the pivot terms  $\alpha_k$  are positive numbers, thus the quadratic form is positive definite, which was to be demonstrated.

# PROBLEMS

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## PROBLEM 1

Diagonalize the matrix

$$A = \begin{pmatrix} 0 & 1 & 1 \\ 1 & 0 & 1 \\ 1 & 1 & 0 \end{pmatrix}$$

## PROBLEM 2

Diagonalize the matrix

$$A = \begin{pmatrix} 4 & -3 & -3 \\ 3 & -2 & -3 \\ -1 & 1 & 2 \end{pmatrix}$$

Compute the matrix  $A^{10}$ .



## PROBLEM 3

Diagonalize the matrix

$$A = \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix}$$

## PROBLEM 4

Consider the matrix

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

Compute  $A^{10}$ .

Write  $A^{10}$  as a linear combination of dyads.

## PROBLEM 5

Diagonalize the matrix

$$A = \begin{pmatrix} 0 & 1 & 0 \\ -1 & 0 & 0 \\ 0 & 0 & 2 \end{pmatrix}$$

## PROBLEM 6

Diagonalize the matrix

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

# TRACE OF A MATRIX

## Definition

If  $A$  is a square matrix, then the sum of the diagonal terms of  $A$  is called the *trace* of matrix  $A$ . Notation:  $\text{tr } A$ .

Prove that the trace of similar matrices are the same. That is for any regular matrix  $P$

$$\text{tr}(P^{-1}AP) = \text{tr } A.$$

Hint: Let us prove first, that  $\text{tr}(AB) = \text{tr}(BA)$  for any square matrices  $A$  and  $B$ .

## PROBLEM 7

Diagonalize the matrix

$$A = \begin{pmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{pmatrix}$$

## PROBLEM 8

Diagonalize the matrix

$$A = \begin{pmatrix} 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 1 \\ 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 1 \end{pmatrix}$$

# UPPER TRIANGULAR MATRICES

Prove that an upper triangular matrix is regular if and only if the diagonal terms are non zero numbers.

Prove that  $\lambda$  is an eigenvalue of an upper triangular matrix if and only if  $\lambda$  is one of the terms of the diagonal of the matrix.