

Advanced Mathematical Methods

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1 Linear Algebra

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WIRTSCHAFTS- UND
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Outline: Linear Algebra

1.8 Eigenvalues and eigenvectors

1.9 Quadratic forms and sign definiteness

Readings

- ▶ Knut Sydsaeter, Peter Hammond, Atle Seierstad, and Arne Strøm. *Further Mathematics for Economic Analysis*. Prentice Hall, 2008 Chapter 1

Online Resources

MIT course on Linear Algebra (by Gilbert Strang)

- ▶ Lecture 21: Eigenvalues and Eigenvectors
<https://www.youtube.com/watch?v=IXNXrLcoerU>
- ▶ Lecture 22: Powers of a square matrix and Diagonalization
<https://www.youtube.com/watch?v=13r9QY6cmjc>
- ▶ Lecture 26: Symmetric matrices and positive definiteness
<https://www.youtube.com/watch?v=umt6BB1nJ4w>
- ▶ Lecture 27: Positive definite matrices and minima – Quadratic forms
<https://www.youtube.com/watch?v=vF7eyJ2g3kU>

1.8 Eigenvalues and eigenvectors

assume a scalar λ exists such that

$$\mathbf{Ax} = \lambda\mathbf{x}$$

λ : eigenvalue

\mathbf{x} : eigenvector

find λ via the homogenous linear equation system

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

1.8 Eigenvalues and eigenvectors

The properties of a quadratic homogenous linear equation system imply that:

- ▶ in any case a solution does exist;
- ▶ if $\det(\mathbf{A} - \lambda \mathbf{I}) \neq 0$, then $\bar{\mathbf{x}} = \mathbf{0}$ is the trivial solution;
- ▶ only if $\det(\mathbf{A} - \lambda \mathbf{I}) = 0$ there is a non-trivial solution.

1.8 Eigenvalues and eigenvectors

Determination of the eigenvalues via *characteristic equation*:

$$|\mathbf{A} - \lambda \mathbf{I}| = 0 \iff (-1)^n \lambda^n + \alpha_{n-1} \lambda^{n-1} + \dots + \alpha_1 \lambda + \alpha_0 = 0$$

for every (real or complex) eigenvalue λ_i of the $(n \times n)$ -Matrix \mathbf{A} we can calculate the respective eigenvector $\mathbf{x}_i \neq \mathbf{0}$ solving the homogenous linear equation system

$$(\mathbf{A} - \lambda_i \mathbf{I})\mathbf{x}_i = \mathbf{0}. \tag{1}$$

The properties of homogenous linear equation systems imply that the solution of eq. (1) is not unambiguous, i.e. for the eigenvalue λ_i we can find infinitely many eigenvectors \mathbf{x}_i

1.8 Eigenvalues and eigenvectors

\mathbf{A} und \mathbf{B} (quadratic matrices of order n) are similar if a regular $(n \times n)$ - matrix \mathbf{C} exists, such that

$$\mathbf{B} = \mathbf{C}^{-1}\mathbf{A}\mathbf{C} .$$

Special case: **symmetric matrices**

For a symmetric $(n \times n)$ -matrix \mathbf{A} it holds that the normalized eigenvectors $\tilde{\mathbf{x}}_j$ with $j = 1, \dots, n$ have the property

- (1) $\tilde{\mathbf{x}}_j' \tilde{\mathbf{x}}_j = 1$ for all j and
- (2) $\tilde{\mathbf{x}}_i' \tilde{\mathbf{x}}_j = 0$ for all $i \neq j$.

1.8 Eigenvalues and eigenvectors

Principle axis theorem

collecting the normalized eigenvectors $\tilde{\mathbf{x}}_j$ ($j = 1, \dots, n$) in a new matrix $\mathbf{T} = [\tilde{\mathbf{x}}_1 \cdots \tilde{\mathbf{x}}_n]$ with the property $\mathbf{T}^{-1} = \mathbf{T}'$ yields the diagonalization of \mathbf{A} as follows:

$$\mathbf{D} = \mathbf{T}'\mathbf{A}\mathbf{T} = \mathbf{T}^{-1}\mathbf{A}\mathbf{T} = \begin{bmatrix} \lambda_1 & 0 & \dots & 0 \\ 0 & \lambda_2 & \dots & \vdots \\ \vdots & & \ddots & 0 \\ 0 & \dots & 0 & \lambda_n \end{bmatrix}$$

1.8 Eigenvalues and eigenvectors

Properties of eigenvalues

- 1) The product of the eigenvalues of a $(n \times n)$ -matrix yields its determinant: $|\mathbf{A}| = \prod_{i=1}^n \lambda_i$.
- 2) From 1.) it follows that a singular matrix must have at least one eigenvalue $\lambda_j = 0$.
- 3) The matrices \mathbf{A} and \mathbf{A}' have the same eigenvalues.
- 4) For a non-singular matrix \mathbf{A} with eigenvalues λ we have:
 $|\mathbf{A}^{-1} - \frac{1}{\lambda} \mathbf{I}| = 0$.
- 5) Symmetric matrices have only real eigenvalues.
- 6) The rank of a symmetric matrix \mathbf{A} is equal to the number of eigenvalues different from zero.

1.9 Quadratic forms and sign definiteness

Definitions

- ▶ Degree of a polynomial
- ▶ Form of n th degree
- ▶ special case: quadratic form

$$Q(x_1, x_2) = a_{11}x_1^2 + 2a_{12}x_1x_2 + a_{22}x_2^2$$

1.9 Quadratic forms and sign definiteness

A quadratic form $Q(x_1, x_2)$ for two variables x_1 and x_2 is defined as

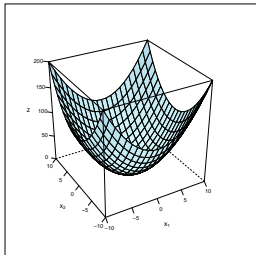
$$Q(x_1, x_2) = \underset{(1 \times 2)}{\mathbf{x}'} \underset{(2 \times 2)}{\mathbf{A}} \underset{(2 \times 1)}{\mathbf{x}} = \sum_{i=1}^2 \sum_{j=1}^2 a_{ij} x_i x_j$$

where $a_{ij} = a_{ji}$ and, thus,

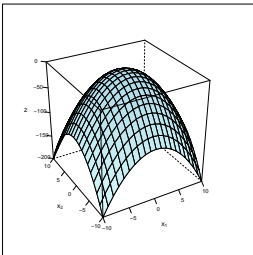
with the symmetric coefficient matrix $\mathbf{A} = \begin{bmatrix} a_{11} & a_{12} \\ a_{12} & a_{22} \end{bmatrix}$

1.9 Quadratic forms and sign definiteness

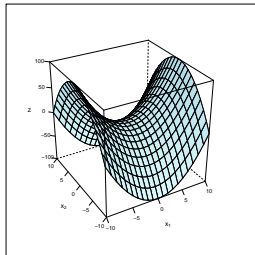
Graph of the positive definite form $Q(x_1, x_2) = x_1^2 + x_2^2$



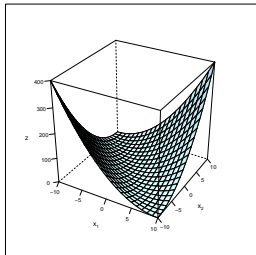
Graph of the negative definite form $Q(x_1, x_2) = -x_1^2 - x_2^2$



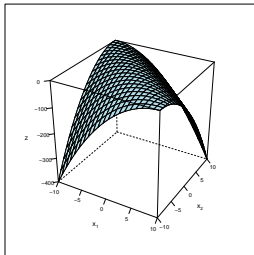
Graph of the indefinite form $Q(x_1, x_2) = x_1^2 - x_2^2$



Graph of the positive semidefinite form $Q(x_1, x_2) = (x_1 + x_2)^2$



Graph of the negative semidefinite form $Q(x_1, x_2) = -(x_1 + x_2)^2$



1.9 Quadratic forms and sign definiteness

The quadratic form associated with the matrix \mathbf{A} (and thus the matrix \mathbf{A} itself) is said to be

positive definite,	if $Q = \mathbf{x}'\mathbf{A}\mathbf{x} > 0$	for all $\mathbf{x} \neq \mathbf{0}$
positive semi-definite,	if $Q = \mathbf{x}'\mathbf{A}\mathbf{x} \geq 0$	for all \mathbf{x}
negative definite,	if $Q = \mathbf{x}'\mathbf{A}\mathbf{x} < 0$	for all $\mathbf{x} \neq \mathbf{0}$
negative semi-definite,	if $Q = \mathbf{x}'\mathbf{A}\mathbf{x} \leq 0$	for all \mathbf{x}

Otherwise the quadratic form is **indefinite**.

Note: For any quadratic matrix \mathbf{A} it holds that $\mathbf{x}'\mathbf{A}\mathbf{x} = \mathbf{x}'\mathbf{B}\mathbf{x}$ with $\mathbf{B} = 0,5 \cdot (\mathbf{A} + \mathbf{A}')$ symmetric.

1.9 Quadratic forms and sign definiteness

The quadratic form $Q(\mathbf{x})$ is

- ▶ positive (negative) definite, if **all** eigenvalues of the matrix \mathbf{A} are positive (negative): $\lambda_j > 0$ ($\lambda_j < 0$) $\forall j = 1, 2, \dots, n$;
- ▶ positive (negative) semi-definite, if **all** eigenvalues of the matrix \mathbf{A} are non-negative (non-positive): $\lambda_j \geq 0$ ($\lambda_j \leq 0$) $\forall j = 1, 2, \dots, n$ and **at least one** eigenvalue is equal to zero;
- ▶ indefinite, if two eigenvalues have different signs.

1.9 Quadratic forms and sign definiteness

Properties of positive definite and positive semi-definite matrices

- 1) Diagonal elements of a positive definite matrix are strictly positive. Diagonal elements of a positive semi-definite matrix are nonnegative.
- 2) If \mathbf{A} is positive definite, then \mathbf{A}^{-1} exists and is positive definite.
- 3) If \mathbf{X} is $n \times k$, then $\mathbf{X}'\mathbf{X}$ and $\mathbf{X}\mathbf{X}'$ are positive semi-definite.
- 4) If \mathbf{X} is $n \times k$ and $\text{rk}(\mathbf{X}) = k$, then $\mathbf{X}'\mathbf{X}$ is positive definite (and therefore non-singular).