

## Definition

Assume  $f : \Omega \rightarrow \mathbb{R}$ ,  $\Omega \rightarrow \mathbb{R}^n$  and  $x_0 \in \Omega$ . If all second order partial derivatives of  $f$  exist at  $x_0$ , then the following matrix

$$\nabla^2 f(x_0) = \begin{pmatrix} \frac{\partial^2 f}{\partial x_1^2}(x_0) & \frac{\partial^2 f}{\partial x_1 \partial x_2}(x_0) & \cdots & \frac{\partial^2 f}{\partial x_1 \partial x_n}(x_0) \\ \frac{\partial^2 f}{\partial x_2 \partial x_1}(x_0) & \frac{\partial^2 f}{\partial x_2^2}(x_0) & \cdots & \frac{\partial^2 f}{\partial x_2 \partial x_n}(x_0) \\ \vdots & & & \\ \frac{\partial^2 f}{\partial x_n \partial x_1}(x_0) & \frac{\partial^2 f}{\partial x_n \partial x_2}(x_0) & \cdots & \frac{\partial^2 f}{\partial x_n^2}(x_0) \end{pmatrix}.$$

is called the Hessian matrix of  $f$  at  $x_0$ .

If second order partial derivatives of  $f$  are all continuous at  $x_0$  then Hessian matrix is symmetric.

### Example

Let  $f : \mathbb{R}^3 \rightarrow \mathbb{R}$  be defined by  $f(x_1, x_2, x_3) = x_1^3 + 4x_2^2x_1 + \sin(x_3)$ .  
Compute the Hessian matrix  $\nabla^2 f(x_0)$  at  $x_0 = (1, 1, 0)^T$ .

## Definition

Assume  $A = [a_{ij}]_{i,j=1}^n$  is a  $n \times n$  symmetric matrix, i.e.  $a_{ij} = a_{ji}$ . A function  $QF_A : \mathbb{R}^n \rightarrow \mathbb{R}$  is called a **quadratic form** associated to the matrix  $A$  if

$$QF_A(h) = h^T A h = \sum_{i=1}^n \sum_{j=1}^n a_{ij} h_i h_j.$$

### Example

Construct the quadratic form associated to the matrix  $A$  if

$$A = \begin{pmatrix} 1 & 3 & 1 \\ 3 & 5 & 0 \\ 1 & 0 & 2 \end{pmatrix}.$$

## Definition

We will say that the symmetric matrix  $A$  or the quadratic form  $QF_A$  is

- **positive definite** if  $QF_A(h) > 0$ ,  $\forall h \in \mathbb{R}^n$  and  $h \neq 0$ ;
- **positive semidefinite** if  $QF_A(h) \geq 0$ ,  $\forall h \in \mathbb{R}^n$ ;
- **negative definite** if  $QF_A(h) < 0$ ,  $\forall h \in \mathbb{R}^n$  and  $h \neq 0$ ;
- **negative semidefinite** if  $QF_A(h) \leq 0$ ,  $\forall h \in \mathbb{R}^n$ ;
- **indefinite** if there exist  $h_1, h_2 \in \mathbb{R}^n$  such that  $QF_A(h_1) > 0$  and  $QF_A(h_2) < 0$ .

## Example

Determine whether the matrix  $A$  is positive definite (semidefinite), negative definite (semidefinite) or indefinite if

a.

$$A = \begin{pmatrix} 1 & 2 & 0 \\ 2 & 5 & 0 \\ 0 & 0 & 2 \end{pmatrix};$$

b.

$$A = \begin{pmatrix} -1 & 3 \\ 3 & -9 \end{pmatrix};$$

c.

$$A = \begin{pmatrix} 1 & 3 \\ 3 & 5 \end{pmatrix}.$$

## Definition

Let  $A = [a_{ij}]_{i,j=1}^n$  be  $n \times n$  matrix. The leading principal minors are  $\det A$  and the minors obtained by successively removing the last row and the last column. That is, the leading principal minors are

$$\Delta_1 = a_{11}, \Delta_2 = \det \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix}, \Delta_3 = \det \begin{pmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{pmatrix}, \dots,$$

$$\Delta_n = \det A.$$

## Theorem

*Let  $A = [a_{ij}]_{i,j=1}^n$  be  $n \times n$  symmetric matrix. The following three statements are equivalent*

- *A is positive definite;*
- *All eigenvalues of A are positive;*
- *All leading principal minors of A are positive (Sylvester's criterion).*



## Definition

Let  $A = [a_{ij}]_{i,j=1}^n$  be  $n \times n$  matrix. The principal minors are  $\det A$  itself and the determinants of matrices obtained by successively removing an  $i$ -th row and  $i$ -th column. That is, the principal minors are

$$\det \begin{pmatrix} a_{i_1 i_1} & a_{i_1 i_2} & \cdots & a_{i_1 i_p} \\ a_{i_2 i_1} & a_{i_2 i_2} & \cdots & a_{i_2 i_p} \\ \vdots & \vdots & \ddots & \vdots \\ a_{i_p i_1} & a_{i_p i_2} & \cdots & a_{i_p i_p} \end{pmatrix}, 1 \leq i_1 < i_2 < \cdots < i_p \leq n, p = 1, \dots, n.$$

## Theorem

*Let  $A = [a_{ij}]_{i,j=1}^n$  be  $n \times n$  symmetric matrix. The following three statements are equivalent*

- *A is positive semidefinite;*
- *All eigenvalues of A are nonnegative;*
- *All principal minors of A are nonnegative.*

## Example

Determine whether the matrix  $A$  is positive definite (semidefinite) if

**a.**

$$A = \begin{pmatrix} 1 & 2 & 0 \\ 2 & 5 & 0 \\ 0 & 0 & 2 \end{pmatrix};$$

**b.**

$$A = \begin{pmatrix} 1 & 3 & 0 \\ 3 & 9 & 0 \\ 0 & 0 & 2 \end{pmatrix}.$$

## Theorem

Let  $A = [a_{ij}]_{i,j=1}^n$  be  $n \times n$  symmetric matrix. The following three statements are equivalent

- $A$  is negative definite;
- All eigenvalues of  $A$  are negative;
- All leading principal minors of even order are positive and of odd order negative (Sylvester's criterion).

## Theorem

*Let  $A = [a_{ij}]_{i,j=1}^n$  be  $n \times n$  symmetric matrix. The following three statements are equivalent*

- *A is negative semidefinite;*
- *All eigenvalues of A are nonpositive;*
- *All principal minors of even order are nonnegative and of odd order nonpositive.*

## Example

Determine whether the matrix  $A$  is positive definite (semidefinite), negative definite (semidefinite) or indefinite if

a.

$$A = \begin{pmatrix} -1 & -2 & 0 \\ -2 & -5 & 0 \\ 0 & 0 & -2 \end{pmatrix};$$

b.

$$A = \begin{pmatrix} -1 & -3 & 0 \\ -3 & -9 & 0 \\ 0 & 0 & -2 \end{pmatrix};$$

c.

$$A = \begin{pmatrix} 1 & 3 & 0 \\ 3 & -9 & 0 \\ 0 & 0 & 2 \end{pmatrix}.$$

## Theorem

*If  $\Omega \subset \mathbb{R}^n$  is an open, convex set and  $f \in \mathbb{C}^2(\Omega)$ , then  $f$  is convex if and only if*

$$\nabla^2 f(x) \succeq 0, \quad \forall x \in \Omega.$$

## Theorem

*If  $\Omega \subset \mathbb{R}^n$  is an open, convex set and  $f \in \mathbb{C}^2(\Omega)$  such that  $\nabla^2 f(x) \succ 0$ ,  $\forall x \in \Omega$ , then  $f$  is strictly convex.*

## Theorem

*If  $\Omega \subset \mathbb{R}^n$  is an open, convex set and  $f \in \mathcal{C}^2(\Omega)$ , then  $f$  is concave if and only if*

$$\nabla^2 f(x) \preceq 0, \quad \forall x \in \Omega.$$

## Theorem

*If  $\Omega \subset \mathbb{R}^n$  is an open, convex set and  $f \in \mathcal{C}^2(\Omega)$  such that  $\nabla^2 f(x) \prec 0$ ,  $\forall x \in \Omega$ , then  $f$  is strictly concave.*



## Example

Check whether  $f$  is convex (strictly convex), concave (strictly concave) on  $\Omega$  if

**a.**  $f(x_1, x_2, x_3) = x_1^2 + 3x_1x_2 + 4x_2^2 + x_3^2 - x_1x_3, \Omega = \mathbb{R}^3;$

**b.**  $f(x_1, x_2) = x_1^2 - 2x_1x_2 - 2x_2^2 + 4x_2, \Omega = \mathbb{R}^2;$

**c.**  $f(x_1, x_2) = -x_1^4 + 2x_1x_2 - x_2^4 - x_1^2 - 2x_2^2, \Omega = \mathbb{R}^2.$