

**MAU23203—Analysis in Several Variables**  
**School of Mathematics, Trinity College**  
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**Section 10: Second Order Partial**  
**Derivatives and the Hessian Matrix**

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### 10.1. Second Order Partial Derivatives

Let  $X$  be an open subset of  $\mathbb{R}^n$  and let  $f: X \rightarrow \mathbb{R}$  be a real-valued function on  $X$ . We consider the second order partial derivatives of the function  $f$  defined by

$$\frac{\partial^2 f}{\partial x_i \partial x_j} = \frac{\partial}{\partial x_i} \left( \frac{\partial f}{\partial x_j} \right).$$

We shall show that if the partial derivatives

$$\frac{\partial f}{\partial x_i}, \quad \frac{\partial f}{\partial x_j}, \quad \frac{\partial^2 f}{\partial x_i \partial x_j} \text{ and } \frac{\partial^2 f}{\partial x_j \partial x_i}$$

all exist and are continuous then

$$\frac{\partial^2 f}{\partial x_i \partial x_j} = \frac{\partial^2 f}{\partial x_j \partial x_i}.$$

Now it would be incorrect to assert that if the second order partial derivatives of a real-valued function  $f$  of real variables  $x_1, x_2, \dots, x_n$  all exist at some point of the domain of the function then

$$\frac{\partial^2 f}{\partial x_i \partial x_j} \quad \text{and} \quad \frac{\partial^2 f}{\partial x_j \partial x_i}.$$

are equal for all values of  $i$  and  $j$ .

## 10. Second Order Partial Derivatives and the Hessian Matrix (continued)

A standard counterexample is provided by the function  $f: \mathbb{R}^2 \rightarrow \mathbb{R}$  that is defined so that

$$f(x, y) = \begin{cases} \frac{xy(x^2 - y^2)}{x^2 + y^2} & \text{if } (x, y) \neq (0, 0); \\ 0 & \text{if } (x, y) = (0, 0). \end{cases}$$

Calculations applying the basic definitions and the standard rules of differential calculus show that the second order partial derivatives of this function  $f$  at every point of its domain  $\mathbb{R}^2$ . However the second order partial derivatives are not continuous at the point  $(0, 0)$ , and moreover

$$\frac{\partial^2 f}{\partial x \partial y} = 1 \quad \text{and} \quad \frac{\partial^2 f}{\partial y \partial x} = -1.$$

**Theorem 10.1**

*Let  $X$  be an open set in  $\mathbb{R}^2$  and let  $f: X \rightarrow \mathbb{R}$  be a real-valued function on  $X$ . Suppose that the partial derivatives*

$$\frac{\partial f}{\partial x}, \quad \frac{\partial f}{\partial y} \quad \text{and} \quad \frac{\partial^2 f}{\partial x \partial y}$$

*exist and are continuous throughout  $X$ . Then the partial derivative*

$$\frac{\partial^2 f}{\partial y \partial x}$$

*exists and is continuous on  $X$ , and*

$$\frac{\partial^2 f}{\partial x \partial y} = \frac{\partial^2 f}{\partial y \partial x}.$$

**Proof**

Let

$$f_x(x, y) = \frac{\partial f}{\partial x}, \quad f_y(x, y) = \frac{\partial f}{\partial y},$$

$$f_{xy}(x, y) = \frac{\partial^2 f}{\partial x \partial y} \text{ and } f_{yx}(x, y) = \frac{\partial^2 f}{\partial y \partial x}$$

and let  $(a, b)$  be a point of  $X$ . The set  $X$  is open in  $\mathbb{R}^2$  and therefore there exists some positive real number  $L$  such that  $(a + h, b + k) \in X$  for all  $(h, k) \in \mathbb{R}^2$  satisfying  $|h| < L$  and  $|k| < L$ .

Let

$$S(h, k) = f(a + h, b + k) + f(a, b) - f(a + h, b) - f(a, b + k)$$

for all real numbers  $h$  and  $k$  satisfying  $|h| < L$  and  $|k| < L$ . First consider  $h$  to be fixed, where  $|h| < L$ , and let  $q: (b - L, b + L) \rightarrow \mathbb{R}$  be defined so that  $q(t) = f(a + h, t) - f(a, t)$  for all real numbers  $t$  satisfying  $b - L < t < b + L$ . Then

$S(h, k) = q(b + k) - q(b)$ . It then follows from the Mean Value Theorem (Theorem 7.5) that there exists some real number  $v$  lying between  $b$  and  $b + k$  for which  $q(b + k) - q(b) = kq'(v)$ . But  $q'(v) = f_y(a + h, v) - f_y(a, v)$ . It follows that

$$S(h, k) = k(f_y(a + h, v) - f_y(a, v)).$$

The Mean Value Theorem can now be applied to the function sending real numbers  $s$  in the interval  $(a - L, a + L)$  to  $f_y(s, v)$  to deduce the existence of a real number  $u$  lying between  $a$  and  $a + h$  for which

$$\begin{aligned} S(h, k) &= k(f_y(a + h, v) - f_y(a, v)) \\ &= hkf_{xy}(u, v) \\ &= hk \left. \frac{\partial^2 f}{\partial x \partial y} \right|_{(x,y)=(u,v)}. \end{aligned}$$



## 10. Second Order Partial Derivatives and the Hessian Matrix (continued)

Now let some positive real number  $\varepsilon$  be given. The function  $f_{xy}$  is continuous. Therefore there exists some real number  $\delta$  satisfying  $0 < \delta < L$  such that  $|f_{xy}(a+h, b+k) - f_{xy}(a, b)| \leq \varepsilon$  whenever  $|h| < \delta$  and  $|k| < \delta$ . It follows that

$$\left| \frac{S(h, k)}{hk} - f_{xy}(a, b) \right| \leq \varepsilon$$

for all real numbers  $h$  and  $k$  satisfying  $0 < |h| < \delta$  and  $0 < |k| < \delta$ . Now

$$\begin{aligned} \lim_{h \rightarrow 0} \frac{S(h, k)}{hk} &= \frac{1}{k} \lim_{h \rightarrow 0} \frac{f(a+h, b+k) - f(a, b+k)}{h} \\ &\quad - \frac{1}{k} \lim_{h \rightarrow 0} \frac{f(a+h, b) - f(a, b)}{h} \\ &= \frac{f_x(a, b+k) - f_x(a, b)}{k}. \end{aligned}$$

## 10. Second Order Partial Derivatives and the Hessian Matrix (continued)

It follows that

$$\left| \frac{f_x(a, b+k) - f_x(a, b)}{k} - f_{xy}(a, b) \right| \leq \varepsilon$$

whenever  $0 < |k| < \delta$ .

Thus the difference quotient  $\frac{f_x(a, b+k) - f_x(a, b)}{k}$  tends to  $f_{xy}(a, b)$  as  $k$  tends to zero, and therefore the second order partial derivative  $f_{yx}$  exists at the point  $(a, b)$  and

$$f_{yx}(a, b) = \lim_{k \rightarrow 0} \frac{f_x(a, b+k) - f_x(a, b)}{k} = f_{xy}(a, b),$$

as required. ■

**Corollary 10.2**

*Let  $X$  be an open set in  $\mathbb{R}^n$  and let  $f: X \rightarrow \mathbb{R}$  be a real-valued function on  $X$ . Suppose that the partial derivatives*

$$\frac{\partial f}{\partial x_i} \text{ and } \frac{\partial^2 f}{\partial x_i \partial x_j}$$

*exist and are continuous on  $X$  for all integers  $i$  and  $j$  between 1 and  $n$ . Then*

$$\frac{\partial^2 f}{\partial x_i \partial x_j} = \frac{\partial^2 f}{\partial x_j \partial x_i}$$

*for all integers  $i$  and  $j$  between 1 and  $n$ .*

## 10.2. Local Maxima and Minima

### Definition

A function  $\varphi: X \rightarrow \mathbb{R}^p$ , defined over an open set  $X$  in  $\mathbb{R}^n$  and mapping that open set into  $\mathbb{R}^p$  for some positive integers  $n$  and  $p$ , is said to be *k times continuously differentiable* if the partial derivatives of the components of the functions  $\varphi$  of all orders less than or equal to  $k$  exist and are continuous throughout the domain  $X$  of the function  $\varphi$ .

## 10. Second Order Partial Derivatives and the Hessian Matrix (continued)

Let  $f: X \rightarrow \mathbb{R}$  be a twice continuously differentiable real-valued function defined over some open subset  $X$  of  $\mathbb{R}^n$ . (In other words, let  $f$  be a real-valued function defined on an open set  $X$  in  $\mathbb{R}^n$  whose first and second order partial derivatives exist and are continuous throughout the domain  $X$  of the function  $f$ .) Suppose that  $f$  has a local minimum at some point  $\mathbf{p}$  of  $X$ , where  $\mathbf{p} = (p_1, p_2, \dots, p_n)$ . Now for each integer  $i$  between 1 and  $n$  the map

$$t \mapsto f(p_1, \dots, p_{i-1}, t, p_{i+1}, \dots, p_n)$$

has a local minimum at  $t = p_i$ . It follows that the derivative of this map vanishes there. Thus if  $f$  has a local minimum at  $\mathbf{p}$  then

$$\left. \frac{\partial f}{\partial x_i} \right|_{\mathbf{x}=\mathbf{p}} = 0.$$

In many situations the values of the second order partial derivatives of a twice continuously differentiable function of several real variables at a stationary point determines the qualitative behaviour of the function around that stationary point, in particular ensuring, in some situations, that the stationary point is a local minimum or a local maximum.

**Proposition 10.3**

*Let  $f$  be a twice continuously differentiable real-valued function defined over an open ball in  $\mathbb{R}^n$  of radius  $\delta$  centred on some point  $\mathbf{p}$  of  $\mathbb{R}^n$ . Then, given any vector  $\mathbf{h}$  in  $\mathbb{R}^n$  satisfying  $|\mathbf{h}| < \delta$ , there exists some real number  $\theta$  satisfying  $0 < \theta < 1$  for which*

$$f(\mathbf{p} + \mathbf{h}) = f(\mathbf{p}) + \sum_{k=1}^n h_k \left. \frac{\partial f}{\partial x_k} \right|_{\mathbf{p}} + \frac{1}{2} \sum_{j,k=1}^n h_j h_k \left. \frac{\partial^2 f}{\partial x_j \partial x_k} \right|_{\mathbf{p} + \theta \mathbf{h}}.$$

**Proof**

Let  $\mathbf{h}$  satisfy  $|\mathbf{h}| < \delta$ , and let  $q(t) = f(\mathbf{p} + t\mathbf{h})$  for all real numbers  $t$  in some appropriately chosen open interval in the real line that contains the real numbers 0 and 1.



## 10. Second Order Partial Derivatives and the Hessian Matrix (continued)

The function  $q$  is the composition function in which the function  $f$  follows the function that sends real numbers  $t$  in the domain of  $q$  to the point  $\mathbf{p} + t\mathbf{h}$  of  $\mathbb{R}^n$ . It follows, on applying the Chain Rule for differentiable functions of several real variables (Theorem 8.20) that

$$q'(t) = \sum_{k=1}^n h_k (\partial_k f)(\mathbf{p} + t\mathbf{h})$$

and

$$q''(t) = \sum_{j,k=1}^n h_j h_k (\partial_j \partial_k f)(\mathbf{p} + t\mathbf{h}),$$

where

$$(\partial_j f)(x_1, x_2, \dots, x_n) = \frac{\partial f(x_1, x_2, \dots, x_n)}{\partial x_j}$$

and

$$(\partial_j \partial_k f)(x_1, x_2, \dots, x_n) = \frac{\partial^2 f(x_1, x_2, \dots, x_n)}{\partial x_j \partial x_k}.$$

## 10. Second Order Partial Derivatives and the Hessian Matrix (continued)

Now

$$q(1) = q(0) + q'(0) + \frac{1}{2}q''(\theta)$$

for some real number  $\theta$  satisfying  $0 < \theta < 1$ . (see Proposition 7.10). Consequently

$$\begin{aligned} f(\mathbf{p} + \mathbf{h}) &= f(\mathbf{p}) + \sum_{k=1}^n h_k (\partial_k f)(\mathbf{p}) \\ &\quad + \frac{1}{2} \sum_{j,k=1}^n h_j h_k (\partial_j \partial_k f)(\mathbf{p} + \theta \mathbf{h}) \\ &= f(\mathbf{p}) + \sum_{k=1}^n h_k \left. \frac{\partial f}{\partial x_k} \right|_{\mathbf{p}} \\ &\quad + \frac{1}{2} \sum_{j,k=1}^n h_j h_k \left. \frac{\partial^2 f}{\partial x_j \partial x_k} \right|_{\mathbf{p} + \theta \mathbf{h}}, \end{aligned}$$

as required. ■

## 10. Second Order Partial Derivatives and the Hessian Matrix (continued)

Let  $f$  be a twice continuously differentiable real-valued function defined over an open ball of radius  $\delta$  about some given point  $\mathbf{p}$  of  $\mathbb{R}^n$ . It follows from Proposition 10.3 that if

$$\left. \frac{\partial f}{\partial x_j} \right|_{\mathbf{p}} = 0$$

for  $j = 1, 2, \dots, n$ , and if  $|\mathbf{h}| < \delta$  then there exists some real number  $\theta$  satisfying  $0 < \theta < 1$  for which

$$f(\mathbf{p} + \mathbf{h}) = f(\mathbf{p}) + \frac{1}{2} \sum_{i=1}^n \sum_{j=1}^n h_i h_j \left. \frac{\partial^2 f}{\partial x_i \partial x_j} \right|_{\mathbf{x}=\mathbf{p}+\theta\mathbf{h}}.$$

## 10. Second Order Partial Derivatives and the Hessian Matrix (continued)

Let  $f$  be a real-valued function defined over an open set in  $\mathbb{R}^n$  whose second order partial derivative are defined at a point  $\mathbf{p}$  of its domain. Let us denote by  $(H_{i,j}(\mathbf{p}))$  the *Hessian matrix* at the point  $\mathbf{p}$ , defined by

$$H_{i,j}(\mathbf{p}) = \left. \frac{\partial^2 f}{\partial x_i \partial x_j} \right|_{\mathbf{x}=\mathbf{p}}.$$

Suppose now that the function  $f$  is twice continuously differentiable on its domain. Then  $H_{i,j}(\mathbf{p}) = H_{j,i}(\mathbf{p})$  for all integers  $i$  and  $j$  between 1 and  $n$ , by Corollary 10.2, and thus the Hessian matrix is symmetric.

## 10. Second Order Partial Derivatives and the Hessian Matrix (continued)

We now recall some facts concerning symmetric matrices.

Let  $(c_{i,j})$  be a symmetric  $n \times n$  matrix.

The matrix  $(c_{i,j})$  is said to be *positive semi-definite* if

$$\sum_{i=1}^n \sum_{j=1}^n c_{i,j} h_i h_j \geq 0 \text{ for all } (h_1, h_2, \dots, h_n) \in \mathbb{R}^n.$$

The matrix  $(c_{i,j})$  is said to be *positive definite* if

$$\sum_{i=1}^n \sum_{j=1}^n c_{i,j} h_i h_j > 0 \text{ for all non-zero } (h_1, h_2, \dots, h_n) \in \mathbb{R}^n.$$

The matrix  $(c_{i,j})$  is said to be *negative semi-definite* if

$$\sum_{i=1}^n \sum_{j=1}^n c_{i,j} h_i h_j \leq 0 \text{ for all } (h_1, h_2, \dots, h_n) \in \mathbb{R}^n.$$

The matrix  $(c_{i,j})$  is said to be *negative definite* if

$$\sum_{i=1}^n \sum_{j=1}^n c_{i,j} h_i h_j < 0 \text{ for all non-zero } (h_1, h_2, \dots, h_n) \in \mathbb{R}^n.$$

The matrix  $(c_{i,j})$  is said to be *indefinite* if it is neither positive semi-definite nor negative semi-definite.

**Lemma 10.4**

*Let  $(c_{i,j})$  be a positive definite symmetric  $n \times n$  matrix. Then there exists some positive real number  $\varepsilon$  that is small enough to ensure that any symmetric  $n \times n$  matrix  $(b_{i,j})$  whose components all satisfy the inequality  $|b_{i,j} - c_{i,j}| < \varepsilon$  is positive definite.*

**Proof**

Let  $S^{n-1}$  be the unit  $(n-1)$ -sphere in  $\mathbb{R}^n$  defined by

$$S^{n-1} = \{(h_1, h_2, \dots, h_n) \in \mathbb{R}^n : h_1^2 + h_2^2 + \dots + h_n^2 = 1\}.$$

Observe that a symmetric  $n \times n$  matrix  $(b_{i,j})$  is positive definite if and only if

$$\sum_{i=1}^n \sum_{j=1}^n b_{i,j} h_i h_j > 0$$

for all  $(h_1, h_2, \dots, h_n) \in S^{n-1}$ . Now the matrix  $(c_{i,j})$  is positive definite, by assumption. Therefore

$$\sum_{i=1}^n \sum_{j=1}^n c_{i,j} h_i h_j > 0$$

for all  $(h_1, h_2, \dots, h_n) \in S^{n-1}$ .



## 10. Second Order Partial Derivatives and the Hessian Matrix (continued)

But  $S^{n-1}$  is a closed bounded set in  $\mathbb{R}^n$ , it therefore follows from Theorem 5.10 that there exists some  $(k_1, k_2, \dots, k_n) \in S^{n-1}$  with the property that

$$\sum_{i=1}^n \sum_{j=1}^n c_{i,j} h_i h_j \geq \sum_{i=1}^n \sum_{j=1}^n c_{i,j} k_i k_j$$

for all  $(h_1, h_2, \dots, h_n) \in S^{n-1}$ . Let

$$A = \sum_{i=1}^n \sum_{j=1}^n c_{i,j} k_i k_j.$$

Then  $A > 0$  and

$$\sum_{i=1}^n \sum_{j=1}^n c_{i,j} h_i h_j \geq A$$

for all  $(h_1, h_2, \dots, h_n) \in S^{n-1}$ . Set  $\varepsilon = A/n^2$ .

## 10. Second Order Partial Derivatives and the Hessian Matrix (continued)

If  $(b_{i,j})$  is a symmetric  $n \times n$  matrix all of whose coefficients satisfy the inequality  $|b_{i,j} - c_{i,j}| < \varepsilon$  then

$$\left| \sum_{i=1}^n \sum_{j=1}^n (b_{i,j} - c_{i,j}) h_i h_j \right| < \varepsilon n^2 = A,$$

for all  $(h_1, h_2, \dots, h_n) \in S^{n-1}$ , hence

$$\sum_{i=1}^n \sum_{j=1}^n b_{i,j} h_i h_j > \sum_{i=1}^n \sum_{j=1}^n c_{i,j} h_i h_j - A \geq 0$$

for all  $(h_1, h_2, \dots, h_n) \in S^{n-1}$ . Thus the matrix  $(b_{i,j})$  is positive definite, as required. ■

Using the fact that a symmetric  $n \times n$  matrix  $(c_{i,j})$  is negative definite if and only if the matrix  $(-c_{i,j})$  is positive definite, we see that if  $(c_{i,j})$  is a negative definite matrix then there exists some  $\varepsilon > 0$  with the following property: if all of the components of a symmetric  $n \times n$  matrix  $(b_{i,j})$  satisfy the inequality  $|b_{i,j} - c_{i,j}| < \varepsilon$  then the matrix  $(b_{i,j})$  is negative definite.

Let  $f: X \rightarrow \mathbb{R}$  be a twice continuously differentiable real-valued function defined over some open set  $X$  in  $\mathbb{R}^n$ , and let  $\mathbf{p}$  be a point of the open set  $X$ . We have already observed that if the function  $f$  has a local maximum or a local minimum at  $\mathbf{p}$  then

$$\left. \frac{\partial f}{\partial x_i} \right|_{\mathbf{x}=\mathbf{p}} = 0 \quad (i = 1, 2, \dots, n).$$

## 10. Second Order Partial Derivatives and the Hessian Matrix (continued)

We now study the behaviour of the function  $f$  around a point  $\mathbf{p}$  at which the first order partial derivatives vanish. We consider the Hessian matrix  $(H_{i,j}(\mathbf{p}))$  defined by

$$H_{i,j}(\mathbf{p}) = \left. \frac{\partial^2 f}{\partial x_i \partial x_j} \right|_{\mathbf{x}=\mathbf{p}}.$$

### Lemma 10.5

*Let  $f: X \rightarrow \mathbb{R}$  be a twice continuously differentiable real-valued function defined over an open set  $X$  in  $\mathbb{R}^n$ , and let  $\mathbf{p}$  be a point of the open set  $X$  at which*

$$\left. \frac{\partial f}{\partial x_i} \right|_{\mathbf{x}=\mathbf{p}} = 0 \quad (i = 1, 2, \dots, n).$$

*If  $f$  has a local minimum at the point  $\mathbf{p}$  then the Hessian matrix  $(H_{i,j}(\mathbf{p}))$  at  $\mathbf{p}$  is positive semi-definite.*

**Proof**

The first order partial derivatives of  $f$  are zero at  $\mathbf{p}$ . It follows that, given any vector  $\mathbf{h} \in \mathbb{R}^n$  which is sufficiently close to  $\mathbf{0}$ , there exists some  $\theta$  satisfying  $0 < \theta < 1$  (where  $\theta$  depends on  $\mathbf{h}$ ) such that

$$f(\mathbf{p} + \mathbf{h}) = f(\mathbf{p}) + \frac{1}{2} \sum_{i=1}^n \sum_{j=1}^n h_i h_j H_{i,j}(\mathbf{p} + \theta \mathbf{h}),$$

where

$$H_{i,j}(\mathbf{p} + \theta \mathbf{h}) = \left. \frac{\partial^2 f}{\partial x_i \partial x_j} \right|_{\mathbf{x}=\mathbf{p}+\theta \mathbf{h}}$$

(see Proposition 10.3).

It follows from this result that

$$\sum_{i=1}^n \sum_{j=1}^n h_i h_j H_{i,j}(\mathbf{p}) = \lim_{t \rightarrow 0} \frac{2(f(\mathbf{p} + t\mathbf{h}) - f(\mathbf{p}))}{t^2} \geq 0.$$

The result follows. ■

Let  $f: X \rightarrow \mathbb{R}$  be a twice continuously differentiable real-valued function defined over some open set in  $\mathbb{R}^n$ , and let  $\mathbf{p}$  be a point of the domain of  $f$  at which the first order partial derivatives of  $f$  are zero. The above lemma shows that if the function  $f$  has a local minimum at  $\mathbf{p}$  then the Hessian matrix of  $f$  is positive semi-definite at  $\mathbf{p}$ . However the fact that the Hessian matrix of  $f$  is positive semi-definite at  $\mathbf{p}$  is not sufficient to ensure that  $f$  has a local minimum at  $\mathbf{p}$ , as the following example shows.



**Example**

Consider the function  $f: \mathbb{R}^2 \rightarrow \mathbb{R}$  defined by  $f(x, y) = x^2 - y^3$ . The first order partial derivatives of  $f$  are zero at  $(0, 0)$ . The Hessian matrix of  $f$  at  $(0, 0)$  is the matrix

$$\begin{pmatrix} 2 & 0 \\ 0 & 0 \end{pmatrix}.$$

This matrix is positive semi-definite. However  $(0, 0)$  is not a local minimum of  $f$  because  $f(0, y) < f(0, 0)$  for all  $y > 0$ .

The following theorem shows that if the Hessian matrix of the function  $f$  is positive definite at a point at which the first order partial derivatives of  $f$  vanish then  $f$  has a local minimum at that point.

### Theorem 10.6

*Let  $f: X \rightarrow \mathbb{R}$  be a twice continuously differentiable real-valued function defined over some open set  $X$  in  $\mathbb{R}^n$ , and let  $\mathbf{p}$  be a point of  $X$  at which*

$$\left. \frac{\partial f}{\partial x_i} \right|_{\mathbf{x}=\mathbf{p}} = 0 \quad (i = 1, 2, \dots, n).$$

*Suppose that the Hessian matrix  $(H_{ij}(\mathbf{p}))$  of the function  $f$  at the point  $\mathbf{p}$  is positive definite. Then  $f$  has a local minimum at  $\mathbf{p}$ .*

**Proof**

The first order partial derivatives of  $f$  take the value zero at  $\mathbf{p}$ . It follows that, given any vector  $\mathbf{h}$  in  $\mathbb{R}^n$  which is sufficiently close to  $\mathbf{0}$ , there exists some  $\theta$  satisfying  $0 < \theta < 1$  (where  $\theta$  depends on  $\mathbf{h}$ ) such that

$$f(\mathbf{p} + \mathbf{h}) = f(\mathbf{p}) + \frac{1}{2} \sum_{i=1}^n \sum_{j=1}^n h_i h_j H_{i,j}(\mathbf{p} + \theta \mathbf{h}),$$

where

$$H_{i,j}(\mathbf{p} + \theta \mathbf{h}) = \left. \frac{\partial^2 f}{\partial x_i \partial x_j} \right|_{\mathbf{x}=\mathbf{p}+\theta \mathbf{h}}$$

(see Proposition 10.3). Suppose that the Hessian matrix  $(H_{i,j}(\mathbf{p}))$  is positive definite. Then there exists some positive real number  $\varepsilon$  small enough to ensure that if  $|H_{i,j}(\mathbf{x}) - H_{i,j}(\mathbf{p})| < \varepsilon$  for all  $i$  and  $j$  then  $(H_{i,j}(\mathbf{x}))$  is positive definite (see Lemma 10.4).

But it follows from the continuity of the second order partial derivatives of  $f$  that there exists some positive real number  $\delta$  small enough to ensure that  $\mathbf{x} \in X$  and  $|H_{i,j}(\mathbf{x}) - H_{i,j}(\mathbf{p})| < \varepsilon$  for all integers  $i$  and  $j$  between 1 and  $n$  whenever  $|\mathbf{x} - \mathbf{p}| < \delta$ . Thus if  $|\mathbf{h}| < \delta$  then  $(H_{i,j}(\mathbf{p} + \theta\mathbf{h}))$  is positive definite for all  $\theta \in (0, 1)$  so that  $f(\mathbf{p} + \mathbf{h}) > f(\mathbf{p})$ . Thus  $\mathbf{p}$  is a local minimum of the function  $f$ . ■

A symmetric  $n \times n$  matrix  $C$  is positive definite if and only if all its eigenvalues are strictly positive. In particular if  $n = 2$  and if  $\lambda_1$  and  $\lambda_2$  are the eigenvalues of a symmetric  $2 \times 2$  matrix  $C$ , then

$$\lambda_1 + \lambda_2 = \text{trace } C, \quad \lambda_1 \lambda_2 = \det C.$$

Thus a symmetric  $2 \times 2$  matrix  $C$  is positive definite if and only if its trace and determinant are both positive.

**Example**

Consider the function  $f: \mathbb{R}^2 \rightarrow \mathbb{R}$  defined by

$$f(x, y) = 4x^2 + 3y^2 - 2xy - x^3 - x^2y - y^3.$$

Now

$$\left. \frac{\partial f(x, y)}{\partial x} \right|_{(x, y) = (0, 0)} = 0 \quad \text{and} \quad \left. \frac{\partial f(x, y)}{\partial y} \right|_{(x, y) = (0, 0)} = 0.$$

The Hessian matrix of  $f$  at  $(0, 0)$  is

$$\begin{pmatrix} 8 & -2 \\ -2 & 6 \end{pmatrix}.$$

The trace and determinant of this matrix are 14 and 44 respectively. Hence this matrix is positive definite. We conclude from Theorem 10.6 that the function  $f$  has a local minimum at  $(0, 0)$ .