

## Higher order partial derivatives

Higher order partial derivatives are defined similarly to higher order derivatives of real functions.

Let  $D \subseteq \mathbb{R}^n$  be an open set and  $f: D \rightarrow \mathbb{R}$  be a real-valued function. Let  $i, j = 1, \dots, n$  and  $a \in D$ . If the  $i$ -th partial derivative exists on  $D$ ,  $\frac{\partial f}{\partial x_i}: D \rightarrow \mathbb{R}$ , we define the  $ij$ -th second order partial derivative of  $f$  at  $a \in D$  as the  $j$ -th partial derivative of  $\frac{\partial f}{\partial x_i}$  at  $a$ , provided this exists. Thus, the  $ij$ -th partial derivative of  $f$  at  $a$

is  $\frac{\partial}{\partial x_j} \left( \frac{\partial f}{\partial x_i} \right) (a)$ . This is usually abbreviated  $\frac{\partial^2 f}{\partial x_j \partial x_i} (a)$  or  $D_j D_i f(a)$ .

If  $i=j$ , then we use the notation  $\frac{\partial^2 f}{\partial x_i^2} (a)$ .

Let now  $k=1, \dots, n$ . Assume that the  $ij$ -th second order partial derivative exists on  $D$ ,  $\frac{\partial^2 f}{\partial x_j \partial x_i}: D \rightarrow \mathbb{R}$ . We define the  $ijk$ -th third order partial derivative of  $f$  at  $a$  as the  $k$ -th partial derivative of  $\frac{\partial^2 f}{\partial x_j \partial x_i}$  at  $a$ , provided that this exists; it is denoted with  $\frac{\partial^3 f}{\partial x_k \partial x_j \partial x_i}$  and if it

exists for all  $a \in D$ , we have a function  $\frac{\partial^3 f}{\partial x_k \partial x_j \partial x_i}: D \rightarrow \mathbb{R}$ .

And so on for still higher order partial derivatives.

When still higher order partial derivatives are in question, certain obvious abbreviations are used. For example

$$\frac{\partial^4 f}{\partial x \partial y^2 \partial z} \text{ means } \frac{\partial}{\partial x} \left( \frac{\partial}{\partial y} \left( \frac{\partial}{\partial y} \left( \frac{\partial f}{\partial z} \right) \right) \right).$$

$f$  is called twice partial differentiable at  $a$  if all second order partial derivatives at  $a$  exist. By induction, for  $k \in \mathbb{N}$ , we say that  $f$  is  $k$ -times partial differentiable at  $a$  if all partial derivatives of order  $k$  at  $a$  exist.

$f$  is called  $k$ -times continuously partial differentiable when  $f$  is  $k$ -times partial differentiable and all the partial derivatives of order  $\leq k$  are continuous.

The large number of possible higher order partial derivatives of a function of several variables is much reduced by the circumstance that the order of performing the partial differentiation is usually irrelevant. The simplest case of this is the equation

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

Theorem 3.54 (of H.A. Schwarz)

Let  $D \subseteq \mathbb{R}^n$  be open and  $f: D \rightarrow \mathbb{R}$ . Assume that  $f$  is twice continuously partial differentiable. Then for all  $a \in D$  and for all  $i, j = 1, \dots, n$

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

Proof

Without loss of generality, we can assume that  $n=2, i=1, j=2$ .  
 Instead of  $(x_1, x_2)$  we write  $(x, y)$ . Thus, we have to prove that

for all  $a = (x_0, y_0) \in D$ ,

$$\frac{\partial^2 f}{\partial y \partial x}(x_0, y_0) = \frac{\partial^2 f}{\partial x \partial y}(x_0, y_0).$$

We consider the sup norm  $\|\cdot\|_\infty$  on  $\mathbb{R}^2$ . Since  $D$  is open, there exists  $r > 0$  s.t.  $U_r((x_0, y_0)) \Rightarrow \{(x, y) \in \mathbb{R}^2 : |x - x_0| < r \text{ and } |y - y_0| < r\} \subseteq D$ .

For any  $(x, y) \in U_r((x_0, y_0))$ , consider the expression

$$S(x, y) = f(x, y) - f(x, y_0) - f(x_0, y) + f(x_0, y_0).$$

For  $y \in \mathbb{R}$  s.t.  $|y - y_0| < r$ , define

$$F_y: ]x_0 - r, x_0 + r[ \rightarrow \mathbb{R}, \quad F_y(x) = f(x, y) - f(x, y_0),$$

so that  $S(x, y) = F_y(x) - F_y(x_0)$ . By the Mean Value Theorem

(Analysis I), there exists  $\eta$  between  $x_0$  and  $x$  s.t.

$$F_y(x) - F_y(x_0) = (x - x_0) \cdot F_y'(\eta) = (x - x_0) \left( \frac{\partial f}{\partial x}(\eta, y) - \frac{\partial f}{\partial x}(\eta, y_0) \right)$$

Let us consider now the function

$$G_\eta: ]y_0 - r, y_0 + r[ \rightarrow \mathbb{R}, \quad G_\eta(y) = \frac{\partial f}{\partial x}(\eta, y)$$

Then we can apply the Mean Value Theorem to this function to get  $\eta$  between  $y_0$  and  $y$  s.t.

$$G_\eta(y) - G_\eta(y_0) = (y - y_0) \cdot G'_\eta(y) = (y - y_0) \frac{\partial^2 f}{\partial y \partial x}(\zeta, \eta).$$

Hence,

$$(1) \quad S(x, y) = (x - x_0) \cdot (y - y_0) \cdot \frac{\partial^2 f}{\partial y \partial x}(\zeta, \eta),$$

where  $\zeta$  is between  $x_0$  and  $x$ ,  $\eta$  is between  $y_0$  and  $y$ .

A similar argument with the roles of the two variables  $x$  and  $y$  reversed gives

$$(2) \quad S(x, y) = (y - y_0)(x - x_0) \frac{\partial^2 f}{\partial x \partial y}(\tilde{\zeta}, \tilde{\eta}),$$

with  $\tilde{\zeta}$  between  $x_0$  and  $x$ ,  $\tilde{\eta}$  between  $y_0$  and  $y$ .

From (1) and (2) we get for  $x \neq x_0, y \neq y_0$  that

$$\frac{\partial^2 f}{\partial y \partial x}(\zeta, \eta) = \frac{\partial^2 f}{\partial x \partial y}(\tilde{\zeta}, \tilde{\eta}).$$

Since  $\frac{\partial^2 f}{\partial y \partial x}$ ,  $\frac{\partial^2 f}{\partial x \partial y}$  are continuous and  $(\zeta, \eta) \rightarrow (x_0, y_0)$ ,

$(\tilde{\zeta}, \tilde{\eta}) \rightarrow (x_0, y_0)$  as  $(x, y) \rightarrow (x_0, y_0)$ , it follows that

$$\begin{aligned} \frac{\partial^2 f}{\partial y \partial x}(x_0, y_0) &= \lim_{(x, y) \rightarrow (x_0, y_0)} \frac{\partial^2 f}{\partial y \partial x}(\zeta, \eta) = \\ &= \lim_{(x, y) \rightarrow (x_0, y_0)} \frac{\partial^2 f}{\partial x \partial y}(\tilde{\zeta}, \tilde{\eta}) = \frac{\partial^2 f}{\partial x \partial y}(x_0, y_0). \quad \square \end{aligned}$$

### Corollary 9.55

Let  $D \subseteq \mathbb{R}^n$  be open and  $f: D \rightarrow \mathbb{R}$  be  $k$ -times continuously partial differentiable. Then

$$\frac{\partial^k f}{\partial x_{i_1} \partial x_{i_2} \dots \partial x_{i_k}} = \frac{\partial^k f}{\partial x_{\pi(k)} \dots \partial x_{\pi(1)}}$$

for any permutation  $\pi$  of  $i_1, \dots, i_k$ .

Proof

By induction on  $k$ , using Theorem 9.54

□

### Definition 9.58

Let  $D \subseteq \mathbb{R}^n$  be open and  $f: D \rightarrow \mathbb{R}$  be twice continuously differentiable. For any  $a \in D$ , the Hessian matrix of  $f$  at  $a$ , denoted by  $(\text{Hess } f)(a)$  is the matrix

$$(\text{Hess } f)(a) = \left( \frac{\partial^2 f}{\partial x_j \partial x_i} \right)_{i,j=1,\dots,n}$$

As a consequence of Theorem 9.54, the Hessian matrix is symmetric.

Example

Let  $f: \mathbb{R}^2 \rightarrow \mathbb{R}$ ,  $f(x_1, x_2) = x_1^2 x_2 + x_1 + x_2^3$ .

Then  $f$  is twice continuously differentiable on  $D$  and for every  $x = (x_1, x_2) \in D$ ,

$$\partial f(x) = \left( \frac{\partial f}{\partial x_1}(x) \quad \frac{\partial f}{\partial x_2}(x) \right) = (2x_1 x_2 + 1 \quad x_1^2 + 3x_2^2),$$

$$\text{grad } f(x) = [\partial f(x)]^T = \begin{pmatrix} 2x_1 x_2 + 1 \\ x_1^2 + 3x_2^2 \end{pmatrix}$$

$$\begin{aligned} \text{Hess } f(x) &= \begin{pmatrix} \frac{\partial^2 f}{\partial x_1^2}(x) & \frac{\partial^2 f}{\partial x_1 \partial x_2}(x) \\ \frac{\partial^2 f}{\partial x_2 \partial x_1}(x) & \frac{\partial^2 f}{\partial x_2^2}(x) \end{pmatrix} \\ &= \begin{pmatrix} 2x_2 & 2x_1 \\ 2x_1 & 6x_2 \end{pmatrix} \end{aligned}$$

## Taylor's Theorem

Our goal in this section is to obtain Taylor approximations for functions of the type  $f: D \rightarrow \mathbb{R}$ , where  $D \subseteq \mathbb{R}^n$  is an open set.

We shall use the following abbreviations. For a multi-index  $\alpha = (\alpha_1, \dots, \alpha_n) \in \mathbb{N}_0^n$  and for  $x = (x_1, \dots, x_n) \in \mathbb{R}^n$ , set

$$|\alpha| := \alpha_1 + \dots + \alpha_n \quad (\text{length of } \alpha)$$

$$\alpha! := \alpha_1! \dots \alpha_n!$$

$$x^\alpha := x_1^{\alpha_1} \dots x_n^{\alpha_n}$$

If  $f$  is  $|\alpha|$ -times continuously partial differentiable, set

$$\partial^\alpha f(x) := \frac{\partial^{|\alpha|} f}{\partial x_1^{\alpha_1} \partial x_2^{\alpha_2} \dots \partial x_n^{\alpha_n}}(x) \quad \text{for every } x \in D,$$

where  $\partial x_i^{\alpha_i} = \underbrace{\partial x_i \dots \partial x_i}_{\alpha_i \text{-times}}$ .

### Lemma 9.57

Let  $D \subseteq \mathbb{R}^n$  be open and  $f: D \rightarrow \mathbb{R}$  be  $k$ -times continuously differentiable. Assume that  $x \in D$  and  $h \in \mathbb{R}^n$  are such that  $[x, x+h] \subseteq D$ .

Then the function

$$f: [0,1] \rightarrow \mathbb{K}, \quad g(t) = f(x+th)$$

is  $k$ -times continuously differentiable and

$$g^{(k)}(t) = \sum_{|\alpha|=k} \frac{k!}{\alpha!} \partial^\alpha f(x+th) \cdot h^\alpha$$

Proof

We prove by induction on  $k$  that

$$(*) \quad g^{(k)}(t) = \sum_{i_1, \dots, i_k=1}^n \frac{\partial^k f}{\partial x_{i_1} \dots \partial x_{i_k}}(x+th) \cdot h_{i_1} \dots h_{i_k}.$$

$k=1$ : Use chain rule (Corollary 3.36) and the fact that

$$g = f \circ \gamma, \quad \text{where } \gamma: [0,1] \rightarrow \mathbb{K}^n, \quad \gamma(t) = x + t h,$$

$$\gamma = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix}, \quad \gamma_i(t) = x_i + t h_i, \quad \text{so } \gamma_i'(t) = h_i.$$

$$\text{It follows that } g'(t) = \sum_{i=1}^n \frac{\partial f}{\partial x_i}(x+th) \cdot h_i.$$

$k \Rightarrow k+1$

$$\begin{aligned} g^{(k+1)}(t) &= \frac{\partial}{\partial t} \left( \sum_{i_1, \dots, i_k=1}^n \frac{\partial^k f}{\partial x_{i_1} \dots \partial x_{i_k}}(x+th) h_{i_1} \dots h_{i_k} \right) \\ &= \sum_{j=1}^n \frac{\partial}{\partial x_j} \left( \sum_{i_1, \dots, i_k=1}^n \frac{\partial^k f}{\partial x_{i_1} \dots \partial x_{i_k}}(x+th) h_{i_1} \dots h_{i_k} \right) h_j \end{aligned}$$

$$= \sum_{i_1, \dots, i_k, i_{k+1}=1}^n \frac{\partial^{k+1} f}{\partial x_{i_{k+1}} \partial x_{i_k} \dots \partial x_{i_1}} (x+th) h_{i_1} \dots h_{i_k} h_{i_{k+1}}.$$

In the following, we shall see how the formula (\*) can be simplified using multi-indices.

To any  $(i_1, \dots, i_k) \in N^k$  we can associate a multi-index  $\alpha = (\alpha_1, \dots, \alpha_n) \in \mathbb{N}_0^n$  by  $\alpha_i :=$  the number of appearances of  $i$  in  $(i_1, \dots, i_k)$ . Then  $|\alpha| = k$  and, by Corollary 9.55,

$$\begin{aligned} \frac{\partial^k f}{\partial x_{i_k} \dots \partial x_{i_1}} (x+th) h_{i_1} \dots h_{i_k} &= \frac{\partial^k f}{\partial x_1^{\alpha_1} \dots \partial x_n^{\alpha_n}} h_1^{\alpha_1} \dots h_n^{\alpha_n} \\ &= \partial^\alpha f(x+th) \cdot h^\alpha. \end{aligned}$$

For any multi-index  $\alpha \in \mathbb{N}_0^n$  with  $|\alpha| = k$  there are exactly  $\frac{k!}{\alpha_1! \dots \alpha_n!}$   $k$ -tuples  $(i_1, \dots, i_k)$  such that  $\alpha$  is the associated multi-index.

It follows that

$$\begin{aligned} g^{(k)}(t) &= \sum_{i_1, \dots, i_k=1}^n \frac{\partial^k f}{\partial x_{i_k} \dots \partial x_{i_1}} (x+th) h_{i_1} \dots h_{i_k} \\ &= \sum_{|\alpha|=k} \frac{k!}{\alpha_1! \dots \alpha_n!} \partial^\alpha f(x+th) \cdot h^\alpha \\ &= \sum_{|\alpha|=k} \frac{k!}{\alpha!} \partial^\alpha f(x+th) \cdot h^\alpha. \end{aligned}$$

□

Theorem 9.58 (Taylor's Theorem)

Let  $D \subseteq \mathbb{R}^n$  be open and  $f: D \rightarrow \mathbb{R}$  be a  $(k+1)$ -times continuously differentiable function. Assume that  $x \in D$  and  $h \in \mathbb{R}^n$  are such that  $[x, x+h] \subseteq D$ .

Then there exists  $\theta \in [0, 1]$  such that

$$f(x+h) = \underbrace{\sum_{|\alpha| \leq k} \frac{\partial^\alpha f(x)}{\alpha!} \cdot h^\alpha}_{\text{the Taylor polynomial of order } k \text{ of } f \text{ at } x} + \sum_{|\alpha|=k+1} \frac{\partial^\alpha f(x+\theta h)}{\alpha!} \cdot h^\alpha.$$

Proof

Let us consider the function  $g: [0, 1] \rightarrow \mathbb{R}$ ,  $g(t) = f(x+th)$ .

By Lemma 9.57,  $g$  is  $(k+1)$ -times continuously differentiable and

$$g^{(m)}(t) = \sum_{|\alpha|=m} \frac{m!}{\alpha!} \partial^\alpha f(x+th) \cdot h^\alpha \quad \text{for all } m=1, \dots, k+1.$$

Applying Taylor's Theorem from Analysis I to  $g$  we get that there exists  $\theta \in [0, 1]$  s.t.

$$\begin{aligned} g(1) &= \sum_{m=0}^k \frac{g^{(m)}(0)}{m!} + \frac{g^{(k+1)}(\theta)}{(k+1)!} \\ &= \sum_{m=0}^k \sum_{|\alpha|=m} \frac{\partial^\alpha f(x)}{\alpha!} \cdot h^\alpha + \sum_{|\alpha|=k+1} \frac{\partial^\alpha f(x+\theta h)}{\alpha!} \cdot h^\alpha \\ &= \sum_{|\alpha| \leq k} \frac{\partial^\alpha f(x)}{\alpha!} \cdot h^\alpha + \sum_{|\alpha|=k+1} \frac{\partial^\alpha f(x+\theta h)}{\alpha!} \cdot h^\alpha. \end{aligned}$$

□

Corollary 9.53

Let  $D \subseteq \mathbb{R}^n$  be open,  $x \in D$  and  $\delta > 0$  be such that  $U_\delta(x) \subseteq D$ .

Assume that  $f: D \rightarrow \mathbb{R}$  is  $k$ -times continuously differentiable.

Then for all  $h \in \mathbb{R}^n$  with  $\|h\| < \delta$ ,

$$f(x+h) = \sum_{|\alpha| \leq k} \frac{\partial^\alpha f(x)}{\alpha!} \cdot h^\alpha + \varphi(h),$$

where  $\varphi: U_\delta(0) \rightarrow \mathbb{R}$  is such that  $\varphi(0) = 0$  and  $\lim_{\substack{h \rightarrow 0 \\ h \neq 0}} \frac{\varphi(h)}{\|h\|^k} = 0$ .

Proof

Let  $h \in \mathbb{R}^n$  with  $\|h\| < \delta$ . Then  $[x, x+h] \subseteq U_\delta(x) \subseteq D$ , so we can apply Taylor's Theorem to get  $\theta \in [0, 1]$  s.t.

$$\begin{aligned} f(x+h) &= \sum_{|\alpha| \leq k-1} \frac{\partial^\alpha f(x)}{\alpha!} h^\alpha + \sum_{|\alpha|=k} \frac{\partial^\alpha f(x+\theta h)}{\alpha!} h^\alpha \\ &= \sum_{|\alpha|=k} \frac{\partial^\alpha f(x)}{\alpha!} h^\alpha + \varphi_2(h), \end{aligned}$$

$$\text{where } \varphi_2(h) := \frac{\partial^\alpha f(x+\theta h) - \partial^\alpha f(x)}{\alpha!}.$$

Since  $\partial^\alpha f$  is continuous, we get that  $\lim_{h \rightarrow 0} \partial^\alpha f(x+\theta h) = \partial^\alpha f(x)$ ,

so  $\lim_{h \rightarrow 0} \frac{\partial^\alpha f(x+\theta h) - \partial^\alpha f(x)}{\alpha!} = 0$ , that is  $\lim_{h \rightarrow 0} \varphi_2(h) = 0$ .

Let us define  $\varphi: U_\delta(0) \rightarrow \mathbb{R}$ ,  $\varphi(h) = \sum_{|\alpha|=k} \varphi_2(h) \cdot h^\alpha$ .

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$$\text{Then } f(x+h) = \sum_{|k| \leq k} \frac{\partial^k f(x)}{k!} \cdot h^k + \varphi(h), \quad \varphi(0) = 0$$

and

$$\begin{aligned} \frac{|\varphi(h)|}{\|h\|^k} &= \frac{\left| \sum_{|k|=k} \varphi_k(a) \cdot h^k \right|}{\|h\|^k} \leq \sum_{|k|=k} |\varphi_k(a)| \cdot \frac{\|h^k\|}{\|h\|^k} \\ &= \sum_{|k|=k} |\varphi_k(a)| \cdot \frac{|h_1|^{k_1} \cdots |h_n|^{k_n}}{\|h\|^{k_1} \cdots \|h\|^{k_n}} \leq \sum_{|k|=k} |\varphi_k(a)|, \end{aligned}$$

since  $|h_i| \leq \|h\|$  for all  $i=1, \dots, n$ .

Since  $\lim_{h \rightarrow 0} \varphi_k(h) = 0$ , it follows that  $\lim_{h \rightarrow 0} \frac{\varphi(h)}{\|h\|^k} = 0$ . □

Let us recall that the canonical scalar product on  $\mathbb{R}^n$  is defined by:

$$\langle x, y \rangle = \sum_{i=1}^n x_i y_i \quad \text{for all } x = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix}, y = \begin{pmatrix} y_1 \\ \vdots \\ y_n \end{pmatrix} \in \mathbb{R}^n.$$

### Corollary 3.60

Let  $D \subseteq \mathbb{R}^n$  be open,  $x \in D$  and  $\delta > 0$  be such that  $U_\delta(x) \subseteq D$ .

Assume that  $f: D \rightarrow \mathbb{R}$  is continuously differentiable.

Then for all  $h \in \mathbb{R}^n$  with  $\|h\| < \delta$ ,

$$f(x+h) = f(x) + \langle \text{grad } f(x), h \rangle + \varphi(h),$$

where  $\varphi: U_\delta(0) \rightarrow \mathbb{R}$  is s.t.  $\varphi(0) = 0$  and  $\lim_{h \rightarrow 0} \frac{\varphi(h)}{\|h\|} = 0$ .

Proof.

Apply Corollary 9.59 with  $h=1$ . We get that

$$f(x+h) = \sum_{|\alpha| \leq 1} \frac{\partial^\alpha f(x)}{\alpha!} \cdot h^\alpha + \varphi(h),$$

where  $\varphi: U_f(0) \rightarrow \mathbb{R}$ ,  $\varphi(0) = 0$  and  $\lim_{h \rightarrow 0} \frac{\varphi(h)}{\|h\|} = 0$ .

We have the following cases:

1)  $|\alpha| = 0$ . The only multi-index  $\alpha \in \mathbb{N}_0^n$  with  $|\alpha| = 0$  is  $\alpha = (0, \dots, 0)$ . It follows that

$$\sum_{|\alpha|=0} \frac{\partial^\alpha f(x)}{\alpha!} \cdot h^\alpha = \frac{\partial^0 f(x)}{0!} \cdot h^0 = f(x).$$

2)  $|\alpha| = 1$ . The multi-indices  $\alpha \in \mathbb{N}_0^n$  with  $|\alpha| = 1$  are the unit vectors

$$e_i = \begin{pmatrix} 0 \\ \vdots \\ 0 \\ 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix} \rightarrow j\text{-th position}$$

It follows that

$$\begin{aligned} \sum_{|\alpha|=1} \frac{\partial^\alpha f(x)}{\alpha!} \cdot h^\alpha &= \sum_{j=1}^n \frac{\partial^{e_j} f(x)}{e_j!} \cdot h^{e_j} \\ &= \sum_{j=1}^n \frac{\partial f}{\partial x_j}(x) \cdot h_j = \langle \text{grad } f(x), h \rangle, \text{ since} \\ \partial^{e_j} f(x) &= \frac{\partial f}{\partial x_j}(x), \quad e_j! = 1 \text{ and } h^{e_j} = h_j. \end{aligned}$$

Thus,  $f(x+h) = f(x) + \langle \text{grad } f(x), h \rangle + \varphi(h)$ .

□

Corollary 3.61

Let  $D \subseteq \mathbb{R}^n$  be open,  $x \in D$  and  $\delta > 0$  be such that  $U_\delta(x) \subseteq D$ .  
Assume that  $f$  is twice continuously differentiable.

Then for all  $h \in \mathbb{R}^n$  with  $\|h\| < \delta$ ,

$$f(x+h) = f(x) + \langle \text{grad } f(x), h \rangle + \frac{1}{2} \langle h, (\text{Hess } f)(x) \cdot h \rangle + \varphi(h),$$

where  $\varphi: U_\delta(x) \rightarrow \mathbb{R}$  is such that  $\varphi(0) = 0$  and  $\lim_{h \rightarrow 0} \frac{\varphi(h)}{\|h\|^2} = 0$ .

Proof

Apply Corollary 3.58 with  $k=2$ . We get that

$$f(x+h) = \sum_{|\alpha| \leq 2} \frac{\partial^\alpha f(x)}{\alpha!} \cdot h^\alpha + \varphi(h),$$

where  $\varphi: U_\delta(x) \rightarrow \mathbb{R}$ ,  $\varphi(0) = 0$  and  $\lim_{h \rightarrow 0} \frac{\varphi(h)}{\|h\|^2} = 0$ .

By the proof of Corollary 3.60, it follows that

$$f(x+h) = f(x) + \langle \text{grad } f(x), h \rangle + \sum_{|\alpha|=2} \frac{\partial^\alpha f(x)}{\alpha!} \cdot h^\alpha + \varphi(h).$$

We have to find out all the multi-indices  $\alpha \in \mathbb{N}_0^n$  with  $|\alpha|=2$ . There are two types of such multi-indices:

$$1) \quad 2e_i = (0, \dots, 0, 2, 0, \dots, 0) \quad 1 \leq i \leq n.$$

$$\partial^{2e_i} f(x) = \frac{\partial^2 f}{\partial x_i^2}(x), \quad (2e_i)! = 2, \quad h^{2e_i} = h_i^2$$

$$2) \quad e_i + e_j = (0, \dots, 1, \dots, 1, \dots, 0) \quad 1 \leq i < j \leq n.$$

$$2^{e_i+e_j} f(x) = \frac{\partial^2 f}{\partial x_i \partial x_j}(x), \quad (e_i+e_j)! = 1, \quad h^{(e_i+e_j)} = h_i h_j.$$

Hence,

$$\sum_{|k|=2} \frac{\partial^k f(x)}{k!} \cdot h^k = \frac{1}{2} \sum_{i=1}^n \frac{\partial^2 f}{\partial x_i^2}(x) \cdot h_i^2 + \sum_{1 \leq i < j \leq n} \frac{\partial^2 f}{\partial x_i \partial x_j}(x) \cdot h_i h_j$$

Since, by Schwarz' Theorem 9.54, we have that

$$\frac{\partial^2 f}{\partial x_i \partial x_j}(x) = \frac{\partial^2 f}{\partial x_j \partial x_i}(x), \text{ it follows that}$$

$$\sum_{1 \leq i < j \leq n} \frac{\partial^2 f}{\partial x_i \partial x_j}(x) \cdot h_i h_j = \frac{1}{2} \sum_{1 \leq i \neq j \leq n} \frac{\partial^2 f}{\partial x_i \partial x_j}(x) h_i h_j.$$

Then

$$\begin{aligned} \sum_{|k|=2} \frac{\partial^k f(x)}{k!} \cdot h^k &= \frac{1}{2} \left( \sum_{i=1}^n \frac{\partial^2 f}{\partial x_i^2}(x) \cdot h_i^2 + \sum_{i \neq j} \frac{\partial^2 f}{\partial x_i \partial x_j}(x) \cdot h_i h_j \right) \\ &= \frac{1}{2} \sum_{i, j=1}^n \frac{\partial^2 f}{\partial x_i \partial x_j}(x) h_i h_j \end{aligned}$$

$$= \frac{1}{2} \langle h, Ah \rangle, \text{ where } A = \text{Hess } f(x).$$

□