

It is of course understood in (14) that  $\mathbf{h} \in R^n$ . If  $|\mathbf{h}|$  is small enough, then  $\mathbf{x} + \mathbf{h} \in E$ , since  $E$  is open. Thus  $\mathbf{f}(\mathbf{x} + \mathbf{h})$  is defined,  $\mathbf{f}(\mathbf{x} + \mathbf{h}) \in R^m$ , and since  $A \in L(R^n, R^m)$ ,  $A\mathbf{h} \in R^m$ . Thus

$$\mathbf{f}(\mathbf{x} + \mathbf{h}) - \mathbf{f}(\mathbf{x}) - A\mathbf{h} \in R^m.$$

The norm in the numerator of (14) is that of  $R^m$ . In the denominator we have the  $R^n$ -norm of  $\mathbf{h}$ .

There is an obvious uniqueness problem which has to be settled before we go any further.

**9.12 Theorem** Suppose  $E$  and  $\mathbf{f}$  are as in Definition 9.11,  $\mathbf{x} \in E$ , and (14) holds with  $A = A_1$  and with  $A = A_2$ . Then  $A_1 = A_2$ .

**Proof** If  $B = A_1 - A_2$ , the inequality

$$|B\mathbf{h}| \leq |\mathbf{f}(\mathbf{x} + \mathbf{h}) - \mathbf{f}(\mathbf{x}) - A_1\mathbf{h}| + |\mathbf{f}(\mathbf{x} + \mathbf{h}) - \mathbf{f}(\mathbf{x}) - A_2\mathbf{h}|$$

shows that  $|B\mathbf{h}|/|\mathbf{h}| \rightarrow 0$  as  $\mathbf{h} \rightarrow \mathbf{0}$ . For fixed  $\mathbf{h} \neq \mathbf{0}$ , it follows that

$$(16) \quad \frac{|B(t\mathbf{h})|}{|t\mathbf{h}|} \rightarrow 0 \quad \text{as } t \rightarrow 0.$$

The linearity of  $B$  shows that the left side of (16) is independent of  $t$ . Thus  $B\mathbf{h} = 0$  for every  $\mathbf{h} \in R^n$ . Hence  $B = 0$ .

### 9.13 Remarks

(a) The relation (14) can be rewritten in the form

$$(17) \quad \mathbf{f}(\mathbf{x} + \mathbf{h}) - \mathbf{f}(\mathbf{x}) = \mathbf{f}'(\mathbf{x})\mathbf{h} + \mathbf{r}(\mathbf{h})$$

where the remainder  $\mathbf{r}(\mathbf{h})$  satisfies

$$(18) \quad \lim_{\mathbf{h} \rightarrow \mathbf{0}} \frac{|\mathbf{r}(\mathbf{h})|}{|\mathbf{h}|} = 0.$$

We may interpret (17), as in Sec. 9.10, by saying that for fixed  $\mathbf{x}$  and small  $\mathbf{h}$ , the left side of (17) is approximately equal to  $\mathbf{f}'(\mathbf{x})\mathbf{h}$ , that is, to the value of a linear transformation applied to  $\mathbf{h}$ .

(b) Suppose  $\mathbf{f}$  and  $E$  are as in Definition 9.11, and  $\mathbf{f}$  is differentiable in  $E$ . For every  $\mathbf{x} \in E$ ,  $\mathbf{f}'(\mathbf{x})$  is then a function, namely, a linear transformation of  $R^n$  into  $R^m$ . But  $\mathbf{f}'$  is also a function:  $\mathbf{f}'$  maps  $E$  into  $L(R^n, R^m)$ .

(c) A glance at (17) shows that  $\mathbf{f}$  is continuous at any point at which  $\mathbf{f}$  is differentiable.

(d) The derivative defined by (14) or (17) is often called the *differential* of  $\mathbf{f}$  at  $\mathbf{x}$ , or the *total derivative* of  $\mathbf{f}$  at  $\mathbf{x}$ , to distinguish it from the partial derivatives that will occur later.

**9.14 Example** We have defined derivatives of functions carrying  $R^n$  to  $R^m$  to be linear transformations of  $R^n$  into  $R^m$ . What is the derivative of such a linear transformation? The answer is very simple.

*If  $A \in L(R^n, R^m)$  and if  $\mathbf{x} \in R^n$ , then*

$$(19) \quad A'(\mathbf{x}) = A.$$

Note that  $\mathbf{x}$  appears on the left side of (19), but not on the right. Both sides of (19) are members of  $L(R^n, R^m)$ , whereas  $A\mathbf{x} \in R^m$ .

The proof of (19) is a triviality, since

$$(20) \quad A(\mathbf{x} + \mathbf{h}) - A\mathbf{x} = A\mathbf{h},$$

by the linearity of  $A$ . With  $\mathbf{f}(\mathbf{x}) = A\mathbf{x}$ , the numerator in (14) is thus 0 for every  $\mathbf{h} \in R^n$ . In (17),  $\mathbf{r}(\mathbf{h}) = 0$ .

We now extend the chain rule (Theorem 5.5) to the present situation.

**9.15 Theorem** Suppose  $E$  is an open set in  $R^n$ ,  $\mathbf{f}$  maps  $E$  into  $R^m$ ,  $\mathbf{f}$  is differentiable at  $\mathbf{x}_0 \in E$ ,  $\mathbf{g}$  maps an open set containing  $\mathbf{f}(E)$  into  $R^k$ , and  $\mathbf{g}$  is differentiable at  $\mathbf{f}(\mathbf{x}_0)$ . Then the mapping  $\mathbf{F}$  of  $E$  into  $R^k$  defined by

$$\mathbf{F}(\mathbf{x}) = \mathbf{g}(\mathbf{f}(\mathbf{x}))$$

is differentiable at  $\mathbf{x}_0$ , and

$$(21) \quad \mathbf{F}'(\mathbf{x}_0) = \mathbf{g}'(\mathbf{f}(\mathbf{x}_0))\mathbf{f}'(\mathbf{x}_0).$$

On the right side of (21), we have the product of two linear transformations, as defined in Sec. 9.6.

**Proof** Put  $\mathbf{y}_0 = \mathbf{f}(\mathbf{x}_0)$ ,  $A = \mathbf{f}'(\mathbf{x}_0)$ ,  $B = \mathbf{g}'(\mathbf{y}_0)$ , and define

$$\mathbf{u}(\mathbf{h}) = \mathbf{f}(\mathbf{x}_0 + \mathbf{h}) - \mathbf{f}(\mathbf{x}_0) - A\mathbf{h},$$

$$\mathbf{v}(\mathbf{k}) = \mathbf{g}(\mathbf{y}_0 + \mathbf{k}) - \mathbf{g}(\mathbf{y}_0) - B\mathbf{k},$$

for all  $\mathbf{h} \in R^n$  and  $\mathbf{k} \in R^m$  for which  $\mathbf{f}(\mathbf{x}_0 + \mathbf{h})$  and  $\mathbf{g}(\mathbf{y}_0 + \mathbf{k})$  are defined.

Then

$$(22) \quad |\mathbf{u}(\mathbf{h})| = \varepsilon(\mathbf{h})|\mathbf{h}|, \quad |\mathbf{v}(\mathbf{k})| = \eta(\mathbf{k})|\mathbf{k}|,$$

where  $\varepsilon(\mathbf{h}) \rightarrow 0$  as  $\mathbf{h} \rightarrow \mathbf{0}$  and  $\eta(\mathbf{k}) \rightarrow 0$  as  $\mathbf{k} \rightarrow \mathbf{0}$ .

Given  $\mathbf{h}$ , put  $\mathbf{k} = \mathbf{f}(\mathbf{x}_0 + \mathbf{h}) - \mathbf{f}(\mathbf{x}_0)$ . Then

$$(23) \quad |\mathbf{k}| = |A\mathbf{h} + \mathbf{u}(\mathbf{h})| \leq [\|A\| + \varepsilon(\mathbf{h})] |\mathbf{h}|,$$

and

$$\begin{aligned} \mathbf{F}(\mathbf{x}_0 + \mathbf{h}) - \mathbf{F}(\mathbf{x}_0) - B A \mathbf{h} &= \mathbf{g}(\mathbf{y}_0 + \mathbf{k}) - \mathbf{g}(\mathbf{y}_0) - B A \mathbf{h} \\ &= B(\mathbf{k} - A\mathbf{h}) + \mathbf{v}(\mathbf{k}) \\ &= B\mathbf{u}(\mathbf{h}) + \mathbf{v}(\mathbf{k}). \end{aligned}$$

Hence (22) and (23) imply, for  $\mathbf{h} \neq \mathbf{0}$ , that

$$\frac{|\mathbf{F}(\mathbf{x}_0 + \mathbf{h}) - \mathbf{F}(\mathbf{x}_0) - B\mathbf{A}\mathbf{h}|}{|\mathbf{h}|} \leq \|B\| \varepsilon(\mathbf{h}) + [\|A\| + \varepsilon(\mathbf{h})]\eta(\mathbf{k}).$$

Let  $\mathbf{h} \rightarrow \mathbf{0}$ . Then  $\varepsilon(\mathbf{h}) \rightarrow 0$ . Also,  $\mathbf{k} \rightarrow \mathbf{0}$ , by (23), so that  $\eta(\mathbf{k}) \rightarrow 0$ . It follows that  $\mathbf{F}'(\mathbf{x}_0) = B\mathbf{A}$ , which is what (21) asserts.

**9.16 Partial derivatives** We again consider a function  $\mathbf{f}$  that maps an open set  $E \subset R^n$  into  $R^m$ . Let  $\{\mathbf{e}_1, \dots, \mathbf{e}_n\}$  and  $\{\mathbf{u}_1, \dots, \mathbf{u}_m\}$  be the standard bases of  $R^n$  and  $R^m$ . The *components of  $\mathbf{f}$*  are the real functions  $f_1, \dots, f_m$  defined by

$$(24) \quad \mathbf{f}(\mathbf{x}) = \sum_{i=1}^m f_i(\mathbf{x})\mathbf{u}_i \quad (\mathbf{x} \in E),$$

or, equivalently, by  $f_i(\mathbf{x}) = \mathbf{f}(\mathbf{x}) \cdot \mathbf{u}_i$ ,  $1 \leq i \leq m$ .

For  $\mathbf{x} \in E$ ,  $1 \leq i \leq m$ ,  $1 \leq j \leq n$ , we define

$$(25) \quad (D_j f_i)(\mathbf{x}) = \lim_{t \rightarrow 0} \frac{f_i(\mathbf{x} + t\mathbf{e}_j) - f_i(\mathbf{x})}{t},$$

provided the limit exists. Writing  $f_i(x_1, \dots, x_n)$  in place of  $f_i(\mathbf{x})$ , we see that  $D_j f_i$  is the derivative of  $f_i$  with respect to  $x_j$ , keeping the other variables fixed. The notation

$$(26) \quad \frac{\partial f_i}{\partial x_j}$$

is therefore often used in place of  $D_j f_i$ , and  $D_j f_i$  is called a *partial derivative*.

In many cases where the existence of a derivative is sufficient when dealing with functions of one variable, continuity or at least boundedness of the partial derivatives is needed for functions of several variables. For example, the functions  $f$  and  $g$  described in Exercise 7, Chap. 4, are not continuous, although their partial derivatives exist at every point of  $R^2$ . Even for continuous functions, the existence of all partial derivatives does not imply differentiability in the sense of Definition 9.11; see Exercises 6 and 14, and Theorem 9.21.

However, if  $\mathbf{f}$  is known to be differentiable at a point  $\mathbf{x}$ , then its partial derivatives exist at  $\mathbf{x}$ , and they determine the linear transformation  $\mathbf{f}'(\mathbf{x})$  completely:

**9.17 Theorem** Suppose  $\mathbf{f}$  maps an open set  $E \subset R^n$  into  $R^m$ , and  $\mathbf{f}$  is differentiable at a point  $\mathbf{x} \in E$ . Then the partial derivatives  $(D_j f_i)(\mathbf{x})$  exist, and

$$(27) \quad \mathbf{f}'(\mathbf{x})\mathbf{e}_j = \sum_{i=1}^m (D_j f_i)(\mathbf{x})\mathbf{u}_i \quad (1 \leq j \leq n).$$

Here, as in Sec. 9.16,  $\{\mathbf{e}_1, \dots, \mathbf{e}_n\}$  and  $\{\mathbf{u}_1, \dots, \mathbf{u}_m\}$  are the standard bases of  $R^n$  and  $R^m$ .

**Proof** Fix  $j$ . Since  $\mathbf{f}$  is differentiable at  $\mathbf{x}$ ,

$$\mathbf{f}(\mathbf{x} + t\mathbf{e}_j) - \mathbf{f}(\mathbf{x}) = \mathbf{f}'(\mathbf{x})(t\mathbf{e}_j) + \mathbf{r}(t\mathbf{e}_j)$$

where  $|\mathbf{r}(t\mathbf{e}_j)|/t \rightarrow 0$  as  $t \rightarrow 0$ . The linearity of  $\mathbf{f}'(\mathbf{x})$  shows therefore that

$$(28) \quad \lim_{t \rightarrow 0} \frac{\mathbf{f}(\mathbf{x} + t\mathbf{e}_j) - \mathbf{f}(\mathbf{x})}{t} = \mathbf{f}'(\mathbf{x})\mathbf{e}_j.$$

If we now represent  $\mathbf{f}$  in terms of its components, as in (24), then (28) becomes

$$(29) \quad \lim_{t \rightarrow 0} \sum_{i=1}^m \frac{f_i(\mathbf{x} + t\mathbf{e}_j) - f_i(\mathbf{x})}{t} \mathbf{u}_i = \mathbf{f}'(\mathbf{x})\mathbf{e}_j.$$

It follows that each quotient in this sum has a limit, as  $t \rightarrow 0$  (see Theorem 4.10), so that each  $(D_j f_i)(\mathbf{x})$  exists, and then (27) follows from (29).

Here are some consequences of Theorem 9.17:

Let  $[\mathbf{f}'(\mathbf{x})]$  be the matrix that represents  $\mathbf{f}'(\mathbf{x})$  with respect to our standard bases, as in Sec. 9.9.

Then  $\mathbf{f}'(\mathbf{x})\mathbf{e}_j$  is the  $j$ th column vector of  $[\mathbf{f}'(\mathbf{x})]$ , and (27) shows therefore that the number  $(D_j f_i)(\mathbf{x})$  occupies the spot in the  $i$ th row and  $j$ th column of  $[\mathbf{f}'(\mathbf{x})]$ . Thus

$$[\mathbf{f}'(\mathbf{x})] = \begin{bmatrix} (D_1 f_1)(\mathbf{x}) & \cdots & (D_n f_1)(\mathbf{x}) \\ \cdots & \cdots & \cdots \\ (D_1 f_m)(\mathbf{x}) & \cdots & (D_n f_m)(\mathbf{x}) \end{bmatrix}.$$

If  $\mathbf{h} = \sum h_j \mathbf{e}_j$  is any vector in  $R^n$ , then (27) implies that

$$(30) \quad \mathbf{f}'(\mathbf{x})\mathbf{h} = \sum_{i=1}^m \left\{ \sum_{j=1}^n (D_j f_i)(\mathbf{x}) h_j \right\} \mathbf{u}_i.$$

**9.18 Example** Let  $\gamma$  be a differentiable mapping of the segment  $(a, b) \subset R^1$  into an open set  $E \subset R^n$ , in other words,  $\gamma$  is a differentiable curve in  $E$ . Let  $f$  be a real-valued differentiable function with domain  $E$ . Thus  $f$  is a differentiable mapping of  $E$  into  $R^1$ . Define

$$(31) \quad g(t) = f(\gamma(t)) \quad (a < t < b).$$

The chain rule asserts then that

$$(32) \quad g'(t) = f'(\gamma(t))\gamma'(t) \quad (a < t < b).$$

Since  $\gamma'(t) \in L(R^1, R^n)$  and  $f'(\gamma(t)) \in L(R^n, R^1)$ , (32) defines  $g'(t)$  as a linear operator on  $R^1$ . This agrees with the fact that  $g$  maps  $(a, b)$  into  $R^1$ . However,  $g'(t)$  can also be regarded as a real number. (This was discussed in Sec. 9.10.) This number can be computed in terms of the partial derivatives of  $f$  and the derivatives of the components of  $\gamma$ , as we shall now see.

With respect to the standard basis  $\{e_1, \dots, e_n\}$  of  $R^n$ ,  $[\gamma'(t)]$  is the  $n$  by 1 matrix (a "column matrix") which has  $\gamma'_i(t)$  in the  $i$ th row, where  $\gamma_1, \dots, \gamma_n$  are the components of  $\gamma$ . For every  $x \in E$ ,  $[f'(x)]$  is the 1 by  $n$  matrix (a "row matrix") which has  $(D_j f)(x)$  in the  $j$ th column. Hence  $[g'(t)]$  is the 1 by 1 matrix whose only entry is the real number

$$(33) \quad g'(t) = \sum_{i=1}^n (D_i f)(\gamma(t)) \gamma'_i(t).$$

This is a frequently encountered special case of the chain rule. It can be rephrased in the following manner.

Associate with each  $x \in E$  a vector, the so-called "gradient" of  $f$  at  $x$ , defined by

$$(34) \quad (\nabla f)(x) = \sum_{i=1}^n (D_i f)(x) e_i.$$

Since

$$(35) \quad \gamma'(t) = \sum_{i=1}^n \gamma'_i(t) e_i,$$

(33) can be written in the form

$$(36) \quad g'(t) = (\nabla f)(\gamma(t)) \cdot \gamma'(t),$$

the scalar product of the vectors  $(\nabla f)(\gamma(t))$  and  $\gamma'(t)$ .

Let us now fix an  $x \in E$ , let  $u \in R^n$  be a unit vector (that is,  $|u| = 1$ ), and specialize  $\gamma$  so that

$$(37) \quad \gamma(t) = x + tu \quad (-\infty < t < \infty).$$

Then  $\gamma'(t) = u$  for every  $t$ . Hence (36) shows that

$$(38) \quad g'(0) = (\nabla f)(x) \cdot u.$$

On the other hand, (37) shows that

$$g(t) - g(0) = f(x + tu) - f(x).$$

Hence (38) gives

$$(39) \quad \lim_{t \rightarrow 0} \frac{f(x + tu) - f(x)}{t} = (\nabla f)(x) \cdot u.$$

The limit in (39) is usually called the *directional derivative* of  $f$  at  $\mathbf{x}$ , in the direction of the unit vector  $\mathbf{u}$ , and may be denoted by  $(D_{\mathbf{u}}f)(\mathbf{x})$ .

If  $f$  and  $\mathbf{x}$  are fixed, but  $\mathbf{u}$  varies, then (39) shows that  $(D_{\mathbf{u}}f)(\mathbf{x})$  attains its maximum when  $\mathbf{u}$  is a positive scalar multiple of  $(\nabla f)(\mathbf{x})$ . [The case  $(\nabla f)(\mathbf{x}) = \mathbf{0}$  should be excluded here.]

If  $\mathbf{u} = \sum u_i \mathbf{e}_i$ , then (39) shows that  $(D_{\mathbf{u}}f)(\mathbf{x})$  can be expressed in terms of the partial derivatives of  $f$  at  $\mathbf{x}$  by the formula

$$(40) \quad (D_{\mathbf{u}}f)(\mathbf{x}) = \sum_{i=1}^n (D_i f)(\mathbf{x}) u_i.$$

Some of these ideas will play a role in the following theorem.

**9.19 Theorem** Suppose  $\mathbf{f}$  maps a convex open set  $E \subset R^n$  into  $R^m$ ,  $\mathbf{f}$  is differentiable in  $E$ , and there is a real number  $M$  such that

$$\|\mathbf{f}'(\mathbf{x})\| \leq M$$

for every  $\mathbf{x} \in E$ . Then

$$|\mathbf{f}(\mathbf{b}) - \mathbf{f}(\mathbf{a})| \leq M|\mathbf{b} - \mathbf{a}|$$

for all  $\mathbf{a} \in E$ ,  $\mathbf{b} \in E$ .

**Proof** Fix  $\mathbf{a} \in E$ ,  $\mathbf{b} \in E$ . Define

$$\gamma(t) = (1-t)\mathbf{a} + t\mathbf{b}$$

for all  $t \in R^1$  such that  $\gamma(t) \in E$ . Since  $E$  is convex,  $\gamma(t) \in E$  if  $0 \leq t \leq 1$ . Put

$$\mathbf{g}(t) = \mathbf{f}(\gamma(t)).$$

Then

$$\mathbf{g}'(t) = \mathbf{f}'(\gamma(t))\gamma'(t) = \mathbf{f}'(\gamma(t))(\mathbf{b} - \mathbf{a}),$$

so that

$$|\mathbf{g}'(t)| \leq \|\mathbf{f}'(\gamma(t))\| |\mathbf{b} - \mathbf{a}| \leq M|\mathbf{b} - \mathbf{a}|$$

for all  $t \in [0, 1]$ . By Theorem 5.19,

$$|\mathbf{g}(1) - \mathbf{g}(0)| \leq M|\mathbf{b} - \mathbf{a}|.$$

But  $\mathbf{g}(0) = \mathbf{f}(\mathbf{a})$  and  $\mathbf{g}(1) = \mathbf{f}(\mathbf{b})$ . This completes the proof.

**Corollary** If, in addition,  $\mathbf{f}'(\mathbf{x}) = \mathbf{0}$  for all  $\mathbf{x} \in E$ , then  $\mathbf{f}$  is constant.

**Proof** To prove this, note that the hypotheses of the theorem hold now with  $M = 0$ .

**9.20 Definition** A differentiable mapping  $\mathbf{f}$  of an open set  $E \subset R^n$  into  $R^m$  is said to be *continuously differentiable* in  $E$  if  $\mathbf{f}'$  is a continuous mapping of  $E$  into  $L(R^n, R^m)$ .

More explicitly, it is required that to every  $\mathbf{x} \in E$  and to every  $\varepsilon > 0$  corresponds a  $\delta > 0$  such that

$$\|\mathbf{f}'(\mathbf{y}) - \mathbf{f}'(\mathbf{x})\| < \varepsilon$$

if  $\mathbf{y} \in E$  and  $|\mathbf{x} - \mathbf{y}| < \delta$ .

If this is so, we also say that  $\mathbf{f}$  is a  $\mathcal{C}'$ -mapping, or that  $\mathbf{f} \in \mathcal{C}'(E)$ .

**9.21 Theorem** Suppose  $\mathbf{f}$  maps an open set  $E \subset R^n$  into  $R^m$ . Then  $\mathbf{f} \in \mathcal{C}'(E)$  if and only if the partial derivatives  $D_j f_i$  exist and are continuous on  $E$  for  $1 \leq i \leq m$ ,  $1 \leq j \leq n$ .

**Proof** Assume first that  $\mathbf{f} \in \mathcal{C}'(E)$ . By (27),

$$(D_j f_i)(\mathbf{x}) = (\mathbf{f}'(\mathbf{x})\mathbf{e}_j) \cdot \mathbf{u}_i$$

for all  $i, j$ , and for all  $\mathbf{x} \in E$ . Hence

$$(D_j f_i)(\mathbf{y}) - (D_j f_i)(\mathbf{x}) = \{[\mathbf{f}'(\mathbf{y}) - \mathbf{f}'(\mathbf{x})]\mathbf{e}_j\} \cdot \mathbf{u}_i$$

and since  $|\mathbf{u}_i| = |\mathbf{e}_j| = 1$ , it follows that

$$\begin{aligned} |(D_j f_i)(\mathbf{y}) - (D_j f_i)(\mathbf{x})| &\leq |[\mathbf{f}'(\mathbf{y}) - \mathbf{f}'(\mathbf{x})]\mathbf{e}_j| \\ &\leq \|\mathbf{f}'(\mathbf{y}) - \mathbf{f}'(\mathbf{x})\|. \end{aligned}$$

Hence  $D_j f_i$  is continuous.

For the converse, it suffices to consider the case  $m = 1$ . (Why?) Fix  $\mathbf{x} \in E$  and  $\varepsilon > 0$ . Since  $E$  is open, there is an open ball  $S \subset E$ , with center at  $\mathbf{x}$  and radius  $r$ , and the continuity of the functions  $D_j f$  shows that  $r$  can be chosen so that

$$(41) \quad |(D_j f)(\mathbf{y}) - (D_j f)(\mathbf{x})| < \frac{\varepsilon}{n} \quad (\mathbf{y} \in S, 1 \leq j \leq n).$$

Suppose  $\mathbf{h} = \sum h_j \mathbf{e}_j$ ,  $|\mathbf{h}| < r$ , put  $\mathbf{v}_0 = \mathbf{0}$ , and  $\mathbf{v}_k = h_1 \mathbf{e}_1 + \cdots + h_k \mathbf{e}_k$ , for  $1 \leq k \leq n$ . Then

$$(42) \quad f(\mathbf{x} + \mathbf{h}) - f(\mathbf{x}) = \sum_{j=1}^n [f(\mathbf{x} + \mathbf{v}_j) - f(\mathbf{x} + \mathbf{v}_{j-1})].$$

Since  $|\mathbf{v}_k| < r$  for  $1 \leq k \leq n$  and since  $S$  is convex, the segments with end points  $\mathbf{x} + \mathbf{v}_{j-1}$  and  $\mathbf{x} + \mathbf{v}_j$  lie in  $S$ . Since  $\mathbf{v}_j = \mathbf{v}_{j-1} + h_j \mathbf{e}_j$ , the mean value theorem (5.10) shows that the  $j$ th summand in (42) is equal to

$$h_j (D_j f)(\mathbf{x} + \mathbf{v}_{j-1} + \theta_j h_j \mathbf{e}_j)$$