

Chapter 6

Differentiation of Maps

6.1 Bounded Linear Maps

Definition 6.1. A map L from a vector space X into a vector space Y is said to be *linear* if $L(cx_1 + x_2) = cL(x_1) + L(x_2)$ for all $x_1, x_2 \in X$ and $c \in \mathbb{R}$. We often write Lx instead of $L(x)$, and the collection of all linear maps from X to Y is denoted by $\mathcal{L}(X, Y)$.

Suppose further that X and Y are normed spaces equipped with norms $\|\cdot\|_X$ and $\|\cdot\|_Y$, respectively. A linear map $L : X \rightarrow Y$ is said to be bounded if

$$\sup_{\|x\|_X=1} \|Lx\|_Y < \infty.$$

The collection of all bounded linear maps from X to Y is denoted by $\mathcal{B}(X, Y)$, and the number $\sup_{\|x\|_X=1} \|Lx\|_Y$ is often denoted by $\|L\|_{\mathcal{B}(X, Y)}$.

Example 6.2. Let $L : \mathbb{R}^n \rightarrow \mathbb{R}^m$ be given by $Lx = Ax$, where A is an $m \times n$ matrix. Then Example 1.138 shows that $\|L\|_{\mathcal{B}(\mathbb{R}^n, \mathbb{R}^m)}$ is the square root of the largest eigenvalue of $A^T A$ which is certainly a finite number. Therefore, any linear transformation from \mathbb{R}^n to \mathbb{R}^m is bounded.

Proposition 6.3. Let $(X, \|\cdot\|_X)$ and $(Y, \|\cdot\|_Y)$ be normed spaces, and $L \in \mathcal{B}(X, Y)$. Then

$$\|L\|_{\mathcal{B}(X, Y)} = \sup_{x \neq 0} \frac{\|Lx\|_Y}{\|x\|_X} = \inf \{M > 0 \mid \|Lx\|_Y \leq M\|x\|_X\}.$$

In particular, the first equality implies that

$$\|Lx\|_Y \leq \|L\|_{\mathcal{B}(X, Y)} \|x\|_X \quad \forall x \in X.$$

Proposition 6.4. *Let $(X, \|\cdot\|_X)$ and $(Y, \|\cdot\|_Y)$ be normed spaces, and $L \in \mathcal{L}(X, Y)$. Then L is continuous on X if and only if $L \in \mathcal{B}(X, Y)$.*

Proof. “ \Rightarrow ” Since L is continuous at $0 \in X$, there exists $\delta > 0$ such that

$$\|Lx\|_Y = \|Lx - L0\|_Y < 1 \quad \text{if } \|x\|_X < \delta.$$

Then $\|L(\frac{\delta}{2}x)\|_Y \leq 1$ if $\|\frac{\delta}{2}x\|_X < \delta$; thus by the properties of norm,

$$\|Lx\|_Y \leq \frac{2}{\delta} \quad \text{if } \|x\|_X < \frac{\delta}{2}.$$

Therefore, $\sup_{\|x\|_X=1} \|Lx\|_Y \leq \frac{2}{\delta}$ which implies that $L \in \mathcal{B}(X, Y)$.

“ \Leftarrow ” If $L \in \mathcal{B}(X, Y)$, then $M = \|L\|_{\mathcal{B}(X, Y)} < \infty$, and

$$\|Lx_1 - Lx_2\|_Y = \|L(x_1 - x_2)\|_Y \leq M\|x_1 - x_2\|_X$$

which shows that L is uniformly continuous on X . □

Proposition 6.5. *Let $(X, \|\cdot\|_X)$ and $(Y, \|\cdot\|_Y)$ be normed spaces. Then $(\mathcal{B}(X, Y), \|\cdot\|_{\mathcal{B}(X, Y)})$ is a normed space. Moreover, if $(Y, \|\cdot\|_Y)$ is a Banach space, so is $(\mathcal{B}(X, Y), \|\cdot\|_{\mathcal{B}(X, Y)})$.*

Proof. That $(\mathcal{B}(X, Y), \|\cdot\|_{\mathcal{B}(X, Y)})$ is a normed space is left as an exercise. Now suppose that $(Y, \|\cdot\|_Y)$ is a Banach space. Let $\{L_k\}_{k=1}^\infty \subseteq \mathcal{B}(X, Y)$ be a Cauchy sequence. Then by Proposition 6.3, for each $x \in X$ we have

$$\|L_k x - L_\ell x\|_Y = \|(L_k - L_\ell)x\|_Y \leq \|L_k - L_\ell\|_{\mathcal{B}(X, Y)} \|x\|_X \rightarrow 0 \quad \text{as } k, \ell \rightarrow \infty.$$

Therefore, $\{L_k x\}_{k=1}^\infty$ is a Cauchy sequence in Y ; thus convergent. Suppose that $\lim_{k \rightarrow \infty} L_k x = y$. We then establish a map $x \mapsto y$ which we denoted by L ; that is, $Lx = y$. Then L is linear since if $x_1, x_2 \in X$ and $c \in \mathbb{R}$,

$$L(cx_1 + x_2) = \lim_{k \rightarrow \infty} L_k(cx_1 + x_2) = \lim_{k \rightarrow \infty} (cL_k x_1 + L_k x_2) = cLx_1 + Lx_2.$$

Moreover, since $\{L_k\}_{k=1}^\infty$ is a Cauchy sequence, $\exists M > 0$ such that $\|L_k\|_{\mathcal{B}(X, Y)} \leq M$ for all $k \in \mathbb{N}$. If $\varepsilon > 0$ is given, for each $x \in X$ there exists $N = N_x > 0$ such that

$$\|L_k x - Lx\|_Y < \varepsilon \quad \forall k \geq N_x.$$

Therefore, for $k \geq N_x$,

$$\|Lx\|_Y < \|L_k x\|_Y + \varepsilon \leq \|L_k\|_{\mathcal{B}(X,Y)} \|x\|_X + \varepsilon \leq M \|x\|_X + \varepsilon$$

which implies that $\sup_{\|x\|_X=1} \|Lx\|_Y \leq M + \varepsilon$; thus $L \in \mathcal{B}(X, Y)$.

Finally, we show that $\lim_{k \rightarrow \infty} \|L_k - L\|_{\mathcal{B}(X,Y)} = 0$. Let $x \in X$ and $\varepsilon > 0$ be given. Since $\{L_k\}_{k=1}^{\infty}$ is a Cauchy sequence, there exists $N > 0$ such that $\|L_k - L_\ell\|_{\mathcal{B}(X,Y)} < \frac{\varepsilon}{2}$ if $k, \ell \geq N$. Then if $k \geq N$,

$$\|L_k x - Lx\|_Y = \lim_{\ell \rightarrow \infty} \|L_k x - L_\ell x\|_Y \leq \limsup_{\ell \rightarrow \infty} \|L_k - L_\ell\|_{\mathcal{B}(X,Y)} \|x\|_X \leq \frac{\varepsilon}{2} \|x\|_X$$

which shows that $\|L_k - L\|_{\mathcal{B}(X,Y)} < \varepsilon$ if $k \geq N$. \square

Proposition 6.6. *Let $(X, \|\cdot\|_X)$, $(Y, \|\cdot\|_Y)$, $(Z, \|\cdot\|_Z)$ be normed spaces, and $L \in \mathcal{B}(X, Y)$, $K \in \mathcal{B}(Y, Z)$. Then $K \circ L \in \mathcal{B}(X, Z)$, and*

$$\|K \circ L\|_{\mathcal{B}(X,Z)} \leq \|K\|_{\mathcal{B}(Y,Z)} \|L\|_{\mathcal{B}(X,Y)}.$$

We often write $K \circ L$ as KL if K and L are linear.

Proof. By the properties of the norm of a bounded linear map,

$$\|K \circ L(x)\|_Z = \|K(Lx)\|_Z \leq \|K\|_{\mathcal{B}(Y,Z)} \|Lx\|_Y \leq \|K\|_{\mathcal{B}(Y,Z)} \|L\|_{\mathcal{B}(X,Y)} \|x\|_X. \quad \square$$

From now on, when the domain X and the target Y of a linear map L is clear, we use $\|L\|$ instead of $\|L\|_{\mathcal{B}(X,Y)}$ to simplify the notation.

Theorem 6.7. *Let $(X, \|\cdot\|_X)$ and $(Y, \|\cdot\|_Y)$ be normed spaces, and X be finite dimensional. Then every linear map from X to Y is bounded; that is, $\mathcal{L}(X, Y) = \mathcal{B}(X, Y)$.*

Proof. Suppose that $\dim(X) = n$. Let $\{e_k\}_{k=1}^n \subseteq X$ be a linearly independent set of vectors. From Example 4.24, every two norms on X are equivalent; thus we only focus on the norm $\|\cdot\|_2$ on X induced by the inner product

$$(e_i, e_j)_X = \delta_{ij} \quad \forall i = 1, \dots, n.$$

Since $\{e_k\}_{k=1}^n$ is a linear independent set of vectors, every $x \in X$ can be expressed as a unique linear combination of e_k 's; that is, for all $x \in X$, $\exists c_1 = c_1(x), \dots, c_n = c_n(x) \in \mathbb{R}$ such that

$$x = c_1 e_1 + \dots + c_n e_n.$$

These coefficients c_k 's in fact are determined by $c_k = (x, e_k)_X$, and, by Example 4.24 and the Cauchy-Schwarz inequality, satisfy

$$|c_k(x)| \leq \|x\|_2 \|e_k\|_2 \leq C \|x\|_X.$$

As a consequence, if L is a linear map from X to Y , then

$$\begin{aligned} \|Lx\|_Y &= \|L(c_1(x)e_1 + \cdots + c_n(x)e_n)\|_Y \leq |c_1(x)| \|Le_1\|_Y + \cdots + |c_n(x)| \|Le_n\|_Y \\ &\leq nC \|x\|_X \max \{ \|Le_1\|_Y, \dots, \|Le_n\|_Y \} \leq M \|x\|_X \end{aligned}$$

for some constant $M > 0$; thus $\|L\|_{\mathcal{B}(X,Y)} \leq M < \infty$ which shows that $L \in \mathcal{B}(X, Y)$. \square

Theorem 6.8. *Let $\text{GL}(n)$ be the set of all invertible linear maps on \mathbb{R}^n ; that is,*

$$\text{GL}(n) = \{L \in \mathcal{L}(\mathbb{R}^n, \mathbb{R}^n) \mid L \text{ is one-to-one (and onto)}\}.$$

1. *If $L \in \text{GL}(n)$ and $K \in \mathcal{B}(\mathbb{R}^n, \mathbb{R}^n)$ satisfying $\|K - L\| \|L^{-1}\| < 1$, then $K \in \text{GL}(n)$.*
2. *$\text{GL}(n)$ is an open set of $\mathcal{B}(\mathbb{R}^n, \mathbb{R}^n)$.*
3. *The mapping $L \mapsto L^{-1}$ is continuous on $\text{GL}(n)$.*

Proof. 1. Let $\|L^{-1}\| = \frac{1}{\alpha}$ and $\|K - L\| = \beta$. Then $\beta < \alpha$; thus for every $x \in \mathbb{R}^n$,

$$\begin{aligned} \alpha \|x\|_{\mathbb{R}^n} &= \alpha \|L^{-1}Lx\|_{\mathbb{R}^n} \leq \alpha \|L^{-1}\| \|Lx\|_{\mathbb{R}^n} = \|Lx\|_{\mathbb{R}^n} \leq \|(L - K)x\|_{\mathbb{R}^n} + \|Kx\|_{\mathbb{R}^n} \\ &\leq \beta \|x\|_{\mathbb{R}^n} + \|Kx\|_{\mathbb{R}^n}. \end{aligned}$$

As a consequence, $(\alpha - \beta)\|x\|_{\mathbb{R}^n} \leq \|Kx\|_{\mathbb{R}^n}$ and this implies that $K : \mathbb{R}^n \rightarrow \mathbb{R}^n$ is one-to-one hence invertible.

2. By 1, we find that if $\|K - L\| < \frac{1}{\|L^{-1}\|}$, then $K \in \text{GL}(n)$. Then $D(L, \frac{1}{\|L^{-1}\|}) \subseteq \text{GL}(n)$ if $L \in \text{GL}(n)$. Therefore, $\text{GL}(n)$ is open.

3. Let $L \in \text{GL}(n)$ and $\varepsilon > 0$ be given. Choose $\delta = \min \left\{ \frac{1}{2\|L^{-1}\|}, \frac{\varepsilon}{2\|L^{-1}\|^2} \right\}$. If $\|K - L\| < \delta$, then $K \in \text{GL}(n)$. Since $L^{-1} - K^{-1} = K^{-1}(K - L)L^{-1}$, we find that if $\|K - L\| < \delta$,

$$\|K^{-1}\| - \|L^{-1}\| \leq \|K^{-1} - L^{-1}\| \leq \|K^{-1}\| \|K - L\| \|L^{-1}\| < \frac{1}{2} \|K^{-1}\|$$

which implies that $\|K^{-1}\| < 2\|L^{-1}\|$. Therefore, if $\|K - L\| < \delta$,

$$\|L^{-1} - K^{-1}\| \leq \|K^{-1}\| \|K - L\| \|L^{-1}\| < 2\|L^{-1}\|^2 \delta < \varepsilon. \quad \square$$

Remark 6.9. There is another way to see that $\text{GL}(n)$ is open in $\mathcal{B}(\mathbb{R}^n, \mathbb{R}^n)$. Let $\mathcal{M}(n)$ be the collection of $n \times n$ real matrices, and $\|\cdot\|_2$ be the matrix norm introduced in Example 1.138. Also define $\|\cdot\| : \mathcal{M}(n) \rightarrow \mathbb{R}$ by

$$\|A\| = \max \{|a_{ij}| \mid A = [a_{ij}] 1 \leq i, j \leq n\}.$$

Then $\|\cdot\|$ is also a norm on $\mathcal{M}(n)$. Since $\mathcal{M}(n)$ is finite dimensional (in fact, $\dim \mathcal{M}(n) = n^2$), by Example 4.24 $\|\cdot\|$ and $\|\cdot\|_2$ are equivalent norms on $\mathcal{M}(n)$; that is, there exists $C, c > 0$ such that

$$c\|A\| \leq \|A\|_2 \leq C\|A\| \quad \forall A \in \mathcal{M}(n).$$

Let $\{A_k\}_{k=1}^{\infty} \subseteq \mathcal{M}(n)$ be a sequence of $n \times n$ real matrices. The equivalence between $\|\cdot\|$ and $\|\cdot\|_2$ implies that $A_k \rightarrow A$ in $\mathcal{M}(n)$ if and only if each entry of A_k converges to corresponding entry of A . Therefore, the determinant function is continuous on $\mathcal{M}(n)$. In other words,

$$\lim_{A_k \rightarrow A} \det(A_k) = \det(A) \quad \forall A \in \mathcal{M}(n).$$

Since $\text{GL}(n)$ can be viewed as the collection of $n \times n$ matrices with non-zero determinant; that is,

$$\text{GL}(n) = \{A \in \mathcal{M}(n) \mid \det(A) \neq 0\},$$

by the continuity of the determinant function and Theorem 4.11, we conclude that $\text{GL}(n)$ is open in $\mathcal{M}(n)$.

6.1.1 The matrix representation of linear maps between finite dimensional normed spaces

Let $(X, \|\cdot\|_X)$ and $(Y, \|\cdot\|_Y)$ be two finite dimensional normed spaces. Suppose that $\mathcal{B} = \{e_k\}_{k=1}^n$ and $\tilde{\mathcal{B}} = \{\tilde{e}_k\}_{k=1}^m$ are basis of X and Y , respectively. Then every $x \in X$, and $y \in Y$, there exists unique vectors $(c_1, \dots, c_n) \in \mathbb{R}^n$ and $(d_1, \dots, d_m) \in \mathbb{R}^m$ such that

$$x = c_1 e_1 + \dots + c_n e_n \quad \text{and} \quad y = d_1 \tilde{e}_1 + \dots + d_m \tilde{e}_m.$$

We write $[x]_{\mathcal{B}}$ for the column vector $[c_1, \dots, c_n]^T$ and $[y]_{\tilde{\mathcal{B}}}$ for the column vector $[d_1, \dots, d_m]^T$. Then for each $L \in \mathcal{L}(X, Y)$, the matrix representation of L with respect to basis \mathcal{B} and $\tilde{\mathcal{B}}$, denoted by $[L]_{\mathcal{B}, \tilde{\mathcal{B}}}$, is the matrix $\begin{bmatrix} [Le_1]_{\tilde{\mathcal{B}}} & [Le_2]_{\tilde{\mathcal{B}}} & \dots & [Le_n]_{\tilde{\mathcal{B}}} \end{bmatrix}$. The matrix $[L]_{\mathcal{B}, \tilde{\mathcal{B}}}$ has the property that

$$[Lx]_{\tilde{\mathcal{B}}} = [L]_{\mathcal{B}, \tilde{\mathcal{B}}} [x]_{\mathcal{B}}.$$

6.2 Definition of Derivatives and the Jacobian Matrices

Definition 6.10. Let $(X, \|\cdot\|_X)$ and $(Y, \|\cdot\|_Y)$ be two normed spaces. A map $f : A \subseteq X \rightarrow Y$ is said to be **differentiable** at $x_0 \in A$ if there is a bounded linear map in $\mathcal{B}(X, Y)$, denoted by $(Df)(x_0)$ and called the **derivative** of f at x_0 , such that

$$\lim_{\substack{x \rightarrow x_0 \\ x \in A}} \frac{\|f(x) - f(x_0) - (Df)(x_0)(x - x_0)\|_Y}{\|x - x_0\|_X} = 0,$$

where $(Df)(x_0)(x - x_0)$ denotes the value of the linear map $(Df)(x_0)$ applied to the vector $x - x_0 \in X$ (so $(Df)(x_0)(x - x_0) \in Y$). In other words, f is differentiable at $x_0 \in A$ if there exists $L \in \mathcal{B}(X, Y)$ such that

$$\forall \varepsilon > 0, \exists \delta > 0 \ni \|f(x) - f(x_0) - L(x - x_0)\|_Y \leq \varepsilon \|x - x_0\|_X \text{ whenever } x \in D(x_0, \delta) \cap A.$$

If f is differentiable at each point of A , we say that f is differentiable on A .

Remark 6.11. Suppose that $f : A \rightarrow Y$ is differentiable on A , then Df itself is a map from A to $\mathcal{B}(X, Y)$. For each $x \in A$, $Df(x)$ is a linear map, but Df in general is not linear in x .

Example 6.12. Let $f : (0, \infty) \rightarrow \mathbb{R}$ be given by $f(x) = \frac{1}{x}$. Then f is differentiable at any $x_0 \in (0, \infty)$ since $(Df)(x_0) : \mathbb{R} \rightarrow \mathbb{R}$ is the linear map given by

$$(Df)(x_0)(x) = -\frac{1}{x_0^2} \cdot x.$$

To see this, we observe that

$$\begin{aligned} \lim_{x \rightarrow x_0} \frac{\left| \frac{1}{x} - \frac{1}{x_0} - \frac{-1}{x_0^2}(x - x_0) \right|}{|x - x_0|} &= \lim_{x \rightarrow x_0} \frac{\left| \frac{x_0^2 - xx_0 + x^2 - xx_0}{xx_0^2} \right|}{|x - x_0|} = \lim_{x \rightarrow x_0} \frac{x_0^2 - 2xx_0 + x^2}{xx_0^2|x - x_0|} \\ &= \lim_{x \rightarrow x_0} \frac{|x - x_0|}{xx_0^2} = 0. \end{aligned}$$

Remark 6.13. Let $f : (a, b) \rightarrow \mathbb{R}$ be “differentiable” at $x_0 \in (a, b)$ in the sense of Definition 4.55. The “derivative” $f'(x_0)$ and the derivative $(Df)(x_0)$ is related by $(Df)(x_0)(h) = f'(x_0)h$ since

$$\lim_{x \rightarrow x_0} \frac{|f(x) - f(x_0) - f'(x_0)(x - x_0)|}{|x - x_0|} = 0.$$

Example 6.14. View $(\mathbb{C}, |\cdot|)$ as a normed vector space (over field \mathbb{C}). Then $f : \mathbb{C} \rightarrow \mathbb{R}$ given by $f(z) = |z|^2$ is differentiable at $z_0 = 0$ and $(Df)(0) = 0$ since

$$\lim_{z \rightarrow 0} \frac{||z|^2 - |0|^2 - 0 \cdot (z - 0)|}{|z|} = \lim_{z \rightarrow 0} |z| = 0.$$

However, f is not differentiable at any $z_0 \neq 0$. In fact, in Exercise Problem 6.3 one is asked to show that if $f : \mathbb{C} \rightarrow \mathbb{R}$ is differentiable at z_0 , then $(Df)(z_0) = 0$. Therefore, if f is differentiable at $z_0 \neq 0$, then

$$\begin{aligned} \frac{||z|^2 - |z_0|^2 - 0 \cdot (z - z_0)|}{|z - z_0|} &= \left| \frac{z \cdot \bar{z} - z_0 \cdot \bar{z}_0}{z - z_0} \right| = \left| \frac{(z - z_0) \cdot \bar{z}_0 + z \cdot \overline{(z - z_0)}}{z - z_0} \right| \\ &= \left| \bar{z}_0 + \frac{z \cdot \overline{(z - z_0)}}{z - z_0} \right| = \left| \bar{z}_0 + \overline{z - z_0} + \frac{z_0 \cdot \overline{(z - z_0)}}{z - z_0} \right| \end{aligned}$$

and the limit of the right-hand side as z approaches z_0 does not exist since $\lim_{z \rightarrow z_0} \frac{z_0 \cdot \overline{(z - z_0)}}{z - z_0}$ does not exist (by the fact that the limit as z approaches z_0 from the horizontal and vertical directions are different).

On the other hand, the function $g : \mathbb{R}^2 \rightarrow \mathbb{R}$ given by $g(x, y) \equiv f(x + iy) = x^2 + y^2$ is differentiable everywhere and $(Dg)(a, b)\mathbf{v} = 2av_1 + 2bv_2$ for all $\mathbf{v} = (v_1, v_2) \in \mathbb{R}^2$. To see this,

$$\frac{(a+h)^2 + (b+k)^2 - (a^2 + b^2) - (2ah + 2bk)}{\sqrt{h^2 + k^2}} = \frac{h^2 + k^2}{\sqrt{h^2 + k^2}} \rightarrow 0 \quad \text{as } (h, k) \rightarrow (0, 0)$$

which implies that $\lim_{(x,y) \rightarrow (a,b)} \frac{\|(x^2 + y^2) - (a^2 + b^2) - (Dg)(a, b)(x - a, y - b)\|_{\mathbb{R}^2}}{\|(x - a, y - b)\|_{\mathbb{R}^2}} = 0$.

Example 6.15. Let $(X, \|\cdot\|_X)$ and $(Y, \|\cdot\|_Y)$ be two normed spaces. Then every bounded linear map $L : X \rightarrow Y$ is differentiable. In fact, $(DL)(x_0) = L$ for all $x_0 \in X$ since

$$\lim_{x \rightarrow x_0} \frac{\|Lx - Lx_0 - L(x - x_0)\|_Y}{\|x - x_0\|_X} = 0.$$

Example 6.16. Let $f : \text{GL}(n) \rightarrow \text{GL}(n)$ be given by $f(L) = L^{-1}$, where $\text{GL}(n)$ is defined in Theorem 6.8. Then f is differentiable at any “point” $L \in \text{GL}(n)$ with derivative $(Df)(L) \in \mathcal{B}(\text{GL}(n), \text{GL}(n))$ given by $(Df)(L)(K) = -L^{-1}KL^{-1}$ for all $K \in \text{GL}(n)$. The proof is left as an exercise.

Theorem 6.17. Let $(X, \|\cdot\|_X)$, $(Y, \|\cdot\|_Y)$ be normed vector spaces, $\mathcal{U} \subseteq X$ be an open set, and $f : \mathcal{U} \rightarrow Y$ be differentiable at $x_0 \in \mathcal{U}$. Then $(Df)(x_0)$ is uniquely determined by f .

Proof. Suppose $L_1, L_2 \in \mathcal{B}(X, Y)$ are derivatives of f at x_0 . Let $\varepsilon > 0$ be given and $e \in X$ be a unit vector; that is, $\|e\|_X = 1$. Since \mathcal{U} is open, there exists $r > 0$ such that $D(x_0, r) \subseteq \mathcal{U}$. By Definition 6.10, there exists $0 < \delta < r$ such that

$$\frac{\|f(x) - f(x_0) - L_1(x - x_0)\|_Y}{\|x - x_0\|_X} < \frac{\varepsilon}{2} \quad \text{and} \quad \frac{\|f(x) - f(x_0) - L_2(x - x_0)\|_Y}{\|x - x_0\|_X} < \frac{\varepsilon}{2}$$

if $0 < \|x - x_0\|_X < \delta$. Letting $x = x_0 + \lambda e$ with $0 < |\lambda| < \delta$, we have

$$\begin{aligned} \|L_1 e - L_2 e\|_Y &= \frac{1}{|\lambda|} \|L_1(x - x_0) - L_2(x - x_0)\|_Y \\ &\leq \frac{1}{|\lambda|} (\|f(x) - f(x_0) - L_1(x - x_0)\|_Y + \|f(x) - f(x_0) - L_2(x - x_0)\|_Y) \\ &= \frac{\|f(x) - f(x_0) - L_1(x - x_0)\|_Y}{\|x - x_0\|_X} + \frac{\|f(x) - f(x_0) - L_2(x - x_0)\|_Y}{\|x - x_0\|_X} \\ &< \frac{\varepsilon}{2} + \frac{\varepsilon}{2} = \varepsilon. \end{aligned}$$

Since $\varepsilon > 0$ is arbitrary, we conclude that $L_1 e = L_2 e$ for all unit vectors $e \in X$ which guarantees that $L_1 = L_2$ (since if $x \neq 0$, $L_1 x = \|x\|_X L_1(\frac{x}{\|x\|_X}) = \|x\|_X L_2(\frac{x}{\|x\|_X}) = L_2 x$). \square

Example 6.18. $(Df)(x_0)$ may not be unique if the domain of f is not open. For example, let $A = \{(x, y) \mid 0 \leq x \leq 1, y = 0\}$ be a subset of \mathbb{R}^2 , and $f : A \rightarrow \mathbb{R}$ be given by $f(x, y) = 0$. Fix $x_0 = (a, 0) \in A$, then both of the linear maps

$$L_1(x, y) = 0 \quad \text{and} \quad L_2(x, y) = ay \quad \forall (x, y) \in \mathbb{R}^2$$

satisfy Definition 6.10 since

$$\lim_{(x,0) \rightarrow (a,0)} \frac{|f(x,0) - f(a,0) - L_1(x-a,0)|}{\|(x,0) - (a,0)\|_{\mathbb{R}^2}} = \lim_{(x,0) \rightarrow (a,0)} \frac{|f(x,0) - f(a,0) - L_2(x-a,0)|}{\|(x,0) - (a,0)\|_{\mathbb{R}^2}} = 0.$$

Remark 6.19. Let $\mathcal{U} \subseteq \mathbb{R}^n$ be an open set and suppose that $f : \mathcal{U} \rightarrow \mathbb{R}^m$ is differentiable on \mathcal{U} . Then $Df : \mathcal{U} \rightarrow \mathcal{B}(\mathbb{R}^n, \mathbb{R}^m)$. Treating Df as a map from \mathcal{U} to the normed space $(\mathcal{B}(\mathbb{R}^n, \mathbb{R}^m), \|\cdot\|_{\mathcal{B}(\mathbb{R}^n, \mathbb{R}^m)})$, and suppose that Df is also differentiable on \mathcal{U} . Then the derivative of Df , denoted by $D^2 f$, is a map from \mathcal{U} to $\mathcal{B}(\mathbb{R}^n, \mathcal{B}(\mathbb{R}^n, \mathbb{R}^m))$. In other words, for each $a \in \mathcal{U}$, $(D^2 f)(a) \in \mathcal{B}(\mathbb{R}^n, \mathcal{B}(\mathbb{R}^n, \mathbb{R}^m))$ satisfying

$$\lim_{x \rightarrow a} \frac{\|(Df)(x) - (Df)(a) - (D^2 f)(a)(x - a)\|_{\mathcal{B}(\mathbb{R}^n, \mathbb{R}^m)}}{\|x - a\|_{\mathbb{R}^n}} = 0,$$

here $(D^2 f)(a)$ is bounded linear map from \mathbb{R}^n to $\mathcal{B}(\mathbb{R}^n, \mathbb{R}^m)$; thus $(D^2 f)(a)(x - a) \in \mathcal{B}(\mathbb{R}^n, \mathbb{R}^m)$.

Definition 6.20. Let $\{e_k\}_{k=1}^n$ be the standard basis of \mathbb{R}^n , $\mathcal{U} \subseteq \mathbb{R}^n$ be an open set, $a \in \mathcal{U}$ and $f : \mathcal{U} \rightarrow \mathbb{R}$ be a function. The partial derivative of f at a in the direction e_j , denoted by $\frac{\partial f}{\partial x_j}(a)$, is the limit

$$\lim_{h \rightarrow 0} \frac{f(a + he_j) - f(a)}{h}$$

if it exists. In other words, if $a = (a_1, \dots, a_n)$, then

$$\frac{\partial f}{\partial x_j}(a) = \lim_{h \rightarrow 0} \frac{f(a_1, \dots, a_{j-1}, a_j + h, a_{j+1}, \dots, a_n) - f(a_1, \dots, a_n)}{h}.$$

Theorem 6.21. Suppose $\mathcal{U} \subseteq \mathbb{R}^n$ is an open set and $f : \mathcal{U} \rightarrow \mathbb{R}^m$ is differentiable at $a \in \mathcal{U}$. Then the partial derivatives $\frac{\partial f_i}{\partial x_j}(a)$ exists for all $i = 1, \dots, m$ and $j = 1, \dots, n$, and the matrix representation of the linear map $Df(a)$ with respect to the standard basis of \mathbb{R}^n and \mathbb{R}^m is given by

$$[(Df)(a)] = \begin{bmatrix} \frac{\partial f_1}{\partial x_1}(a) & \cdots & \frac{\partial f_1}{\partial x_n}(a) \\ \vdots & \ddots & \vdots \\ \frac{\partial f_m}{\partial x_1}(a) & \cdots & \frac{\partial f_m}{\partial x_n}(a) \end{bmatrix} \quad \text{or} \quad [(Df)(a)]_{ij} = \frac{\partial f_i}{\partial x_j}(a).$$

Proof. Since \mathcal{U} is open and $a \in \mathcal{U}$, there exists $r > 0$ such that $D(a, r) \subseteq \mathcal{U}$. By the differentiability of f at a , there is $L \in \mathcal{B}(\mathbb{R}^n, \mathbb{R}^m)$ such that for any given $\varepsilon > 0$, there exists $0 < \delta < r$ such that

$$\|f(x) - f(a) - L(x - a)\|_{\mathbb{R}^m} \leq \varepsilon \|x - a\|_{\mathbb{R}^n} \quad \text{whenever } x \in D(a, \delta).$$

In particular, for each $i = 1, \dots, m$,

$$\left| \frac{f_i(a + he_j) - f_i(a)}{h} - (Le_j)_i \right| \leq \left\| \frac{f(a + he_j) - f(a)}{h} - Le_j \right\|_{\mathbb{R}^m} \leq \varepsilon \quad \forall 0 < |h| < \delta, h \in \mathbb{R},$$

where $(Le_j)_i$ denotes the i -th component of Le_j in the standard basis. As a consequence, for each $i = 1, \dots, m$,

$$\lim_{h \rightarrow 0} \frac{f_i(a + he_j) - f_i(a)}{h} = (Le_j)_i \quad \text{exists}$$

and by definition, we must have $(Le_j)_i = \frac{\partial f_i}{\partial x_j}(a)$. Therefore, $L_{ij} = \frac{\partial f_i}{\partial x_j}(a)$. \square

Definition 6.22. Let $\mathcal{U} \subseteq \mathbb{R}^n$ be an open set, and $f : \mathcal{U} \rightarrow \mathbb{R}^m$. The matrix

$$(Jf)(x) \equiv \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \cdots & \frac{\partial f_1}{\partial x_n} \\ \vdots & \ddots & \vdots \\ \frac{\partial f_m}{\partial x_1} & \cdots & \frac{\partial f_m}{\partial x_n} \end{bmatrix} (x) \equiv \begin{bmatrix} \frac{\partial f_1}{\partial x_1}(x) & \cdots & \frac{\partial f_1}{\partial x_n}(x) \\ \vdots & \ddots & \vdots \\ \frac{\partial f_m}{\partial x_1}(x) & \cdots & \frac{\partial f_m}{\partial x_n}(x) \end{bmatrix}$$

is called the **Jacobian matrix** of f at x (if each entry exists). If $n = m$, the determinant of $(Jf)(x)$ is called the **Jacobian** of f at x .

Remark 6.23. A function f might not be differential even if the Jacobian matrix Jf exists; however, if f is differentiable at x_0 , then $(Df)(x)$ can be represented by $(Jf)(x)$; that is, $[(Df)(x)] = (Jf)(x)$.

Example 6.24. Let $f : \mathbb{R}^2 \rightarrow \mathbb{R}^3$ be given by $f(x_1, x_2) = (x_1^2, x_1^3 x_2, x_1^4 x_2^2)$. Suppose that f is differentiable at $x = (x_1, x_2)$, then

$$[(Df)(x)] = \begin{bmatrix} 2x_1 & 0 \\ 3x_1^2 x_2 & x_1^3 \\ 4x_1^3 x_2^2 & 2x_1^4 x_2 \end{bmatrix}.$$

Example 6.25. Let $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ be given by

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

Then $\frac{\partial f}{\partial x}(0, 0) = \frac{\partial f}{\partial y}(0, 0) = 0$; thus if f is differentiable at $(0, 0)$, then $[(Df)(0, 0)] = [0 \ 0]$.

However,

$$\left| f(x, y) - f(0, 0) - [0 \ 0] \begin{bmatrix} x \\ y \end{bmatrix} \right| = \frac{|xy|}{x^2 + y^2} = \frac{|xy|}{(x^2 + y^2)^{\frac{3}{2}}} \sqrt{x^2 + y^2};$$

thus f is not differentiable at $(0, 0)$ since $\frac{|xy|}{(x^2 + y^2)^{\frac{3}{2}}}$ cannot be arbitrarily small even if $x^2 + y^2$ is small.

Example 6.26. Let $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ be given by

$$f(x, y) = \begin{cases} x & \text{if } y = 0, \\ y & \text{if } x = 0, \\ 1 & \text{otherwise.} \end{cases}$$

Then $\frac{\partial f}{\partial x}(0,0) = \lim_{h \rightarrow 0} \frac{f(h,0) - f(0,0)}{h} = \lim_{h \rightarrow 0} \frac{h}{h} = 1$. Similarly, $\frac{\partial f}{\partial y}(0,0) = 1$; thus if f is differentiable at $(0,0)$, then $[(Df)(0,0)] = [1 \ 1]$. However,

$$\left| f(x,y) - f(0,0) - [1 \ 1] \begin{bmatrix} x \\ y \end{bmatrix} \right| = |f(x,y) - (x+y)|;$$

thus if $xy \neq 0$,

$$|f(x,y) - (x+y)| = |1 - x - y| \not\rightarrow 0 \text{ as } (x,y) \rightarrow (0,0), xy \neq 0.$$

Therefore, f is not differentiable at $(0,0)$.

6.3 Conditions for Differentiability

Proposition 6.27. *Let $\mathcal{U} \subseteq \mathbb{R}^n$ be open, $a \in \mathcal{U}$, and $f = (f_1, \dots, f_m) : \mathcal{U} \rightarrow \mathbb{R}^m$. Then f is differentiable at a if and only if f_i is differentiable at a for all $i = 1, \dots, m$. In other words, for vector-valued functions defined on an open subset of \mathbb{R}^n ,*

$$\text{Componentwise differentiable} \Leftrightarrow \text{Differentiable.}$$

Proof. “ \Rightarrow ” Let $(Df)(a)$ be the Jacobian matrix of f at a . Then

$$\forall \varepsilon > 0, \exists \delta > 0 \ni \|f(x) - f(a) - (Df)(a)(x-a)\|_{\mathbb{R}^m} \leq \varepsilon \|x-a\|_{\mathbb{R}^n} \text{ if } \|x-a\|_{\mathbb{R}^n} < \delta.$$

Let $\{e_j\}_{j=1}^m$ be the standard basis of \mathbb{R}^m , and $L_i \in \mathcal{L}(\mathbb{R}^n, \mathbb{R})$ be given by $L_i(h) = e_i^T [(Df)(a)]h$. Then $L_i \in \mathcal{B}(\mathbb{R}^n, \mathbb{R})$ by Theorem 6.7, and if $\|x-a\|_{\mathbb{R}^n} < \delta$,

$$\begin{aligned} |f_i(x) - f_i(a) - L_i(x-a)| &= |e_i \cdot (f(x) - f(a) - (Df)(a)(x-a))| \\ &\leq \|f(x) - f(a) - (Df)(a)(x-a)\|_{\mathbb{R}^m} \leq \varepsilon \|x-a\|_{\mathbb{R}^n}; \end{aligned}$$

thus f_i is differentiable at a with derivatives L_i .

“ \Leftarrow ” Suppose that $f_i : \mathcal{U} \rightarrow \mathbb{R}$ is differentiable at a for each $i = 1, \dots, m$. Then there exists $L_i \in \mathcal{B}(\mathbb{R}^n, \mathbb{R})$ such that

$$\forall \varepsilon > 0, \exists \delta_i > 0 \ni |f_i(x) - f_i(a) - L_i(x-a)| \leq \frac{\varepsilon}{m} \|x-a\|_{\mathbb{R}^n} \text{ if } \|x-a\|_{\mathbb{R}^n} < \delta_i.$$

Let $L \in \mathcal{L}(\mathbb{R}^n, \mathbb{R}^m)$ be given by $Lx = (L_1x, L_2x, \dots, L_mx) \in \mathbb{R}^m$ if $x \in \mathbb{R}^n$. Then $L \in \mathcal{B}(\mathbb{R}^n, \mathbb{R}^m)$ by Theorem 6.7, and

$$\|f(x) - f(a) - L(x-a)\|_{\mathbb{R}^m} \leq \sum_{i=1}^m |f_i(x) - f_i(a) - L_i(x-a)| \leq \varepsilon \|x-a\|_{\mathbb{R}^n}$$

if $\|x-a\|_{\mathbb{R}^n} < \delta = \min \{\delta_1, \dots, \delta_m\}$. □

Theorem 6.28. Let $\mathcal{U} \subseteq \mathbb{R}^n$ be open, $a \in \mathcal{U}$, and $f : \mathcal{U} \rightarrow \mathbb{R}$. If

1. the Jacobian matrix of f exists in a neighborhood of a , and
2. at least $(n - 1)$ entries of the Jacobian matrix of f are continuous at a ,

then f is differentiable at a .

Proof. W.L.O.G. we can assume that $\frac{\partial f}{\partial x_1}, \frac{\partial f}{\partial x_2}, \dots, \frac{\partial f}{\partial x_{n-1}}$ are continuous at a . Let $\{e_j\}_{j=1}^n$ be the standard basis of \mathbb{R}^n , and $\varepsilon > 0$ be given. Since $\frac{\partial f}{\partial x_i}$ is continuous at a for $i = 1, \dots, n - 1$,

$$\exists \delta_i > 0 \ni \left| \frac{\partial f}{\partial x_i}(x) - \frac{\partial f}{\partial x_i}(a) \right| < \frac{\varepsilon}{\sqrt{n}} \text{ whenever } \|x - a\|_{\mathbb{R}^n} < \delta_i.$$

On the other hand, by the definition of the partial derivatives,

$$\exists \delta_n > 0 \ni \left| \frac{f(a + he_n) - f(a)}{h} - \frac{\partial f}{\partial x_n}(a) \right| < \frac{\varepsilon}{\sqrt{n}} \text{ whenever } 0 < |h| < \delta_n.$$

Let $k = x - a$ and $\delta = \min\{\delta_1, \dots, \delta_n\}$. Then

$$\begin{aligned} & \left| f(x) - f(a) - \left[\frac{\partial f}{\partial x_1}(a)(x_1 - a_1) + \dots + \frac{\partial f}{\partial x_n}(a)(x_n - a_n) \right] \right| \\ &= \left| f(a + k) - f(a) - \frac{\partial f}{\partial x_1}(a)k_1 - \dots - \frac{\partial f}{\partial x_n}(a)k_n \right| \\ &= \left| f(a_1 + k_1, \dots, a_n + k_n) - f(a_1, \dots, a_n) - \frac{\partial f}{\partial x_1}(a)k_1 - \dots - \frac{\partial f}{\partial x_n}(a)k_n \right| \\ &\leq \left| f(a_1 + k_1, \dots, a_n + k_n) - f(a_1, a_2 + k_2, \dots, a_n + k_n) - \frac{\partial f}{\partial x_1}(a)k_1 \right| \\ &\quad + \left| f(a_1, a_2 + k_2, \dots, a_n + k_n) - f(a_1, a_2, a_3 + k_3, \dots, a_n + k_n) - \frac{\partial f}{\partial x_2}(a)k_2 \right| \\ &\quad + \dots + \left| f(a_1, \dots, a_{n-1}, a_n + k_n) - f(a_1, \dots, a_n) - \frac{\partial f}{\partial x_n}(a)k_n \right|. \end{aligned}$$

By the mean value theorem,

$$\begin{aligned} & f(a_1, \dots, a_{j-1}, a_j + k_j, \dots, a_n + k_n) - f(a_1, \dots, a_j, a_{j+1} + k_{j+1}, \dots, a_n + k_n) \\ &= k_j \frac{\partial f}{\partial x_j}(a_1, \dots, a_{j-1}, a_j + \theta_j k_j, a_{j+1} + k_{j+1}, \dots, a_n + k_n) \end{aligned}$$

for some $0 < \theta_j < 1$; thus for $j = 1, \dots, n - 1$, if $\|x - a\|_{\mathbb{R}^n} = \|k\|_{\mathbb{R}^n} < \delta$,

$$\begin{aligned} & \left| f(a_1, \dots, a_{j-1}, a_j + k_j, \dots, a_n + k_n) - f(a_1, \dots, a_j, a_{j+1} + k_{j+1}, \dots, a_n + k_n) - \frac{\partial f}{\partial x_j}(a)k_j \right| \\ &= \left| \frac{\partial f}{\partial x_j}(a_1, \dots, a_{j-1}, a_j + \theta_j k_j, a_{j+1} + k_{j+1}, \dots, a_n + k_n) - \frac{\partial f}{\partial x_j}(a) \right| |k_j| \leq \frac{\varepsilon}{\sqrt{n}} |k_j|. \end{aligned}$$

Moreover, if $\|x - a\|_{\mathbb{R}^n} < \delta$, then $|k_n| \leq \|k\|_{\mathbb{R}^n} = \|x - a\|_{\mathbb{R}^n} < \delta \leq \delta_n$; thus

$$\left| f(a_1, \dots, a_{n-1}, a_n + k_n) - f(a_1, \dots, a_n) - \frac{\partial f}{\partial x_n}(a)k_n \right| \leq \frac{\varepsilon}{\sqrt{n}}|k_n|.$$

As a consequence, if $\|x - a\|_{\mathbb{R}^n} < \delta$, by Cauchy's inequality,

$$\begin{aligned} \left| f(x) - f(a) - \left[\frac{\partial f}{\partial x_1}(a)(x_1 - a_1) + \dots + \frac{\partial f}{\partial x_n}(a)(x_n - a_n) \right] \right| \\ \leq \frac{\varepsilon}{\sqrt{n}} \sum_{j=1}^n |k_j| \leq \varepsilon \|k\|_{\mathbb{R}^n} = \varepsilon \|x - a\|_{\mathbb{R}^n} \end{aligned}$$

which implies that f is differentiable at a . □

Remark 6.29. When two or more components of the Jacobian matrix $\left[\frac{\partial f}{\partial x_1} \ \dots \ \frac{\partial f}{\partial x_n} \right]$ of a scalar function f are discontinuous at a point $x_0 \in \mathcal{U}$, in general f is not differentiable at x_0 . For example, both components of the Jacobian matrix of the functions given in Example 6.25, 6.26, 6.41 are discontinuous at $(0, 0)$, and these functions are not differentiable at $(0, 0)$.

Example 6.30. Let $\mathcal{U} = \mathbb{R}^2 \setminus \{(x, 0) \in \mathbb{R}^2 \mid x \geq 0\}$, and $f : \mathcal{U} \rightarrow \mathbb{R}$ be given by

$$f(x, y) = \arg(x + iy) = \begin{cases} \cos^{-1} \frac{x}{\sqrt{x^2 + y^2}} & \text{if } y > 0, \\ \pi & \text{if } y = 0, \\ 2\pi - \cos^{-1} \frac{x}{\sqrt{x^2 + y^2}} & \text{if } y < 0. \end{cases}$$

Then

$$\frac{\partial f}{\partial x}(x, y) = \begin{cases} -\frac{y}{x^2 + y^2} & \text{if } y \neq 0, \\ 0 & \text{if } y = 0, \end{cases} \quad \text{and} \quad \frac{\partial f}{\partial y}(x, y) = \begin{cases} \frac{x}{x^2 + y^2} & \text{if } y \neq 0, \\ \frac{1}{x} & \text{if } y = 0. \end{cases}$$

Since $\frac{\partial f}{\partial x}$ and $\frac{\partial f}{\partial y}$ are both continuous on \mathcal{U} , f is differentiable on \mathcal{U} .

Definition 6.31. Let $\mathcal{U} \subseteq \mathbb{R}^n$ be open, and $f : \mathcal{U} \rightarrow \mathbb{R}^m$ be differentiable on \mathcal{U} . f is said to be **continuously differentiable** on \mathcal{U} if $Df : \mathcal{U} \rightarrow \mathcal{B}(\mathbb{R}^n, \mathbb{R}^m)$ is continuous on \mathcal{U} . The collection of all continuously differentiable mappings from \mathcal{U} to \mathbb{R}^m is denoted by $\mathcal{C}^1(\mathcal{U}; \mathbb{R}^m)$. The collection of all bounded differentiable functions from \mathcal{U} to \mathbb{R}^m whose derivative is continuous and bounded is denoted by $\mathcal{C}_b^1(\mathcal{U}; \mathbb{R}^m)$. In other words,

$$\mathcal{C}^1(\mathcal{U}; \mathbb{R}^m) = \{f : \mathcal{U} \rightarrow \mathbb{R}^m \text{ is differentiable on } \mathcal{U} \mid Df : \mathcal{U} \rightarrow \mathcal{B}(\mathbb{R}^n, \mathbb{R}^m) \text{ is continuous}\}$$

and

$$\mathcal{C}_b^1(\mathcal{U}; \mathbb{R}^m) = \left\{ f \in \mathcal{C}^1(\mathcal{U}; \mathbb{R}^m) \mid \sup_{x \in \mathcal{U}} |f(x)| + \sup_{x \in \mathcal{U}} \|Df(x)\|_{\mathcal{B}(\mathbb{R}^n, \mathbb{R}^m)} < \infty \right\}.$$

Corollary 6.32. *Let $\mathcal{U} \subseteq \mathbb{R}^n$ be open, and $f : \mathcal{U} \rightarrow \mathbb{R}^m$. Then $f \in \mathcal{C}_b^1(\mathcal{U}; \mathbb{R}^m)$ if and only if the partial derivatives $\frac{\partial f_i}{\partial x_j}$ exist and are continuous on \mathcal{U} for $i = 1, \dots, m$ and $j = 1, \dots, n$.*

Proof. Note that for any matrix $A = [a_{ij}]_{m \times n}$, $\|A\|_{\mathcal{B}(\mathbb{R}^n, \mathbb{R}^m)} \leq \sum_{i,j} |a_{ij}| \leq nm\|A\|$; thus

$$\begin{aligned} \|(Df)(x) - (Df)(x_0)\|_{\mathcal{B}(\mathbb{R}^n, \mathbb{R}^m)} &\leq \sum_{i=1}^m \sum_{j=1}^n \left| \frac{\partial f_i}{\partial x_j}(x) - \frac{\partial f_i}{\partial x_j}(x_0) \right| \\ &\leq nm \|(Df)(x) - (Df)(x_0)\|_{\mathcal{B}(\mathbb{R}^n, \mathbb{R}^m)}. \end{aligned}$$

Therefore, the continuity of Df is equivalent to the continuity of the partial derivatives $\frac{\partial f_i}{\partial x_j}$ for all i, j . The corollary is then concluded by Proposition 6.27 and Theorem 6.28. \square

Example 6.33. If $f : \mathbb{R} \rightarrow \mathbb{R}$ is differentiable at x_0 , must f' be continuous at x_0 ? In other words, is it always true that $\lim_{x \rightarrow x_0} f'(x) = f'(x_0)$?

Answer: No! For example, take

$$f(x) = \begin{cases} x^2 \sin \frac{1}{x} & \text{if } x \neq 0, \\ 0 & \text{if } x = 0. \end{cases}$$

Then f is differentiable at $x = 0$ since the limit

$$\lim_{h \rightarrow 0} \frac{f(0+h) - f(0)}{h} = \lim_{h \rightarrow 0} \frac{h^2 \sin \frac{1}{h}}{h} = \lim_{h \rightarrow 0} h \sin \frac{1}{h} = 0$$

exists. Therefore,

$$f'(x) = \begin{cases} 2x \sin \frac{1}{x} - \cos \frac{1}{x} & \text{if } x \neq 0, \\ 0 & \text{if } x = 0. \end{cases}$$

However, $\lim_{x \rightarrow 0} f'(x)$ does not exist.

Proposition 6.34. *Let $\mathcal{U} \subseteq \mathbb{R}^n$ be open. Given $f \in \mathcal{C}_b^1(\mathcal{U}; \mathbb{R}^m)$, define*

$$\|f\|_{\mathcal{C}_b^1(\mathcal{U}; \mathbb{R}^m)} = \sup_{x \in \mathcal{U}} \left[|f(x)| + \sum_{i=1}^m \sum_{j=1}^n \left| \frac{\partial f_i}{\partial x_j}(x) \right| \right].$$

Then $(\mathcal{C}_b^1(\mathcal{U}; \mathbb{R}^m), \|\cdot\|_{\mathcal{C}_b^1(\mathcal{U}; \mathbb{R}^m)})$ is a Banach space.

Proof. Left as an exercise. □

Theorem 6.35. *Let $\mathcal{U} \subseteq \mathbb{R}^n$ be open, $K \subseteq \mathcal{U}$ be compact, and $f : \mathcal{U} \rightarrow \mathbb{R}$ be of class \mathcal{C}^1 . Then for each $\varepsilon > 0$, there exists $\delta > 0$ such that*

$$|f(y) - f(x) - (Df)(x)(y - x)| \leq \varepsilon \|y - x\|_{\mathbb{R}^n} \quad \text{if } \|y - x\|_{\mathbb{R}^n} < \delta \text{ and } x, y \in K.$$

Proof. Define $g : \mathcal{U} \times \mathcal{U} \rightarrow \mathbb{R}$ by

$$g(x, y) = \begin{cases} \frac{|f(y) - f(x) - (Df)(x)(y - x)|}{\|y - x\|_{\mathbb{R}^n}} & \text{if } y \neq x, \\ 0 & \text{if } y = x. \end{cases}$$

Since f is of class \mathcal{C}^1 , g is continuous on $\mathcal{U} \times \mathcal{U}$. In fact, it is clear that g is continuous at (x, y) if $x \neq y$, while the mean value theorem implies that $f(w) - f(z) = (Df)(\xi)(w - z)$ for some ξ on the line segment joining w and z ; thus

$$\begin{aligned} & \limsup_{\substack{(z,w) \rightarrow (x,x) \\ z \neq w}} \frac{|f(w) - f(z) - (Df)(z)(w - z)|}{\|w - z\|_{\mathbb{R}^n}} \\ &= \limsup_{\substack{(z,w) \rightarrow (x,x) \\ z \neq w}} \frac{|((Df)(\xi) - (Df)(z))(w - z)|}{\|w - z\|_{\mathbb{R}^n}} \leq \limsup_{\substack{(z,w) \rightarrow (x,x) \\ z \neq w}} \|(Df)(\xi) - (Df)(z)\|_{\mathcal{B}(\mathbb{R}^n, \mathbb{R})} = 0. \end{aligned}$$

Now by the compactness of $K \times K$, for each given $\varepsilon > 0$, there exists $\delta > 0$ such that

$$|g(z, w) - g(x, y)| < \varepsilon \quad \text{if } \|(z, w) - (x, y)\|_{\mathbb{R}^{2n}} < \delta \text{ and } x, y, z, w \in K.$$

In particular, with $(z, w) = (x, x)$ we find that $|g(x, y)| < \varepsilon$ if $\|x - y\|_{\mathbb{R}^n} < \delta$; thus

$$\frac{|f(y) - f(x) - (Df)(x)(y - x)|}{\|y - x\|_{\mathbb{R}^n}} < \varepsilon \quad \text{if } 0 < \|x - y\|_{\mathbb{R}^n} < \delta, x, y \in K. \quad \square$$

Corollary 6.36. *Let $\mathcal{U} \subseteq \mathbb{R}^n$ be open, $K \subseteq \mathcal{U}$ be compact, and $f : \mathcal{U} \rightarrow \mathbb{R}^m$ be of class \mathcal{C}^1 . Then for each $\varepsilon > 0$, there exists $\delta > 0$ such that*

$$\|f(y) - f(x) - (Df)(x)(y - x)\|_{\mathbb{R}^m} \leq \varepsilon \|y - x\|_{\mathbb{R}^n} \quad \text{if } \|y - x\|_{\mathbb{R}^n} < \delta \text{ and } x, y \in K.$$

6.4 Properties of Differentiable Functions

6.4.1 Continuity of differentiable maps

Theorem 6.37. *Let $(X, \|\cdot\|_X)$ and $(Y, \|\cdot\|_Y)$ be normed spaces, $\mathcal{U} \subseteq X$ be open, and $f : \mathcal{U} \rightarrow Y$ be differentiable at $x_0 \in \mathcal{U}$. Then f is continuous at x_0 .*

Proof. Since f is differentiable at x_0 , there exists $L \in \mathcal{B}(X, Y)$ such that

$$\exists \delta_1 > 0 \ni \|f(x) - f(x_0) - L(x - x_0)\|_Y \leq \|x - x_0\|_X \quad \forall x \in D(x_0, \delta_1).$$

As a consequence,

$$\|f(x) - f(x_0)\|_Y \leq (\|L\| + 1)\|x - x_0\|_X \quad \forall x \in D(x_0, \delta_1). \quad (6.4.1)$$

For a given $\varepsilon > 0$, let $\delta = \min \left\{ \delta_1, \frac{\varepsilon}{2(\|L\| + 1)} \right\}$. Then $\delta > 0$, and if $x \in D(x_0, \delta)$,

$$\|f(x) - f(x_0)\|_Y \leq \frac{\varepsilon}{2} < \varepsilon. \quad \square$$

Remark 6.38. In fact, if f is differentiable at x_0 , then f satisfies the “local Lipschitz property”; that is,

$\exists M = M(x_0) > 0$ and $\delta = \delta(x_0) > 0 \ni$ if $\|x - x_0\|_X < \delta$, then $\|f(x) - f(x_0)\|_Y \leq M\|x - x_0\|_X$ since we can choose $M = \|L\| + 1$ and $\delta = \delta_1$ (see (6.4.1)).

Example 6.39. Let $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ be given in Example 6.25. We have shown that f is not differentiable at $(0, 0)$. In fact, f is not even continuous at $(0, 0)$ since when approaching the origin along the straight line $x_2 = mx_1$,

$$\lim_{(x_1, mx_1) \rightarrow (0, 0)} f(x_1, mx_1) = \lim_{x_1 \rightarrow 0} \frac{mx_1^2}{(m^2 + 1)x_1^2} = \frac{m^2}{m^2 + 1} \neq f(0, 0) \text{ if } m \neq 0.$$

Example 6.40. Let $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ be given in Example 6.26. Then f is not continuous at $(0, 0)$; thus not differentiable at $(0, 0)$.

Example 6.41. Let $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ be given by

$$f(x, y) = \begin{cases} \frac{x^3}{x^2 + y^2} & \text{if } (x, y) \neq (0, 0), \\ 0 & \text{if } (x, y) = (0, 0). \end{cases}$$

Then $f_x(0, 0) = 1$ and $f_y(0, 0) = 0$. However,

$$\frac{\left| f(x, y) - f(0, 0) - \begin{bmatrix} 1 & 0 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} \right|}{\sqrt{x^2 + y^2}} = \frac{|x|y^2}{(x^2 + y^2)^{\frac{3}{2}}} \not\rightarrow 0 \text{ as } (x, y) \rightarrow (0, 0).$$

Therefore, f is not differentiable at $(0, 0)$. On the other hand, f is continuous at $(0, 0)$ since

$$|f(x, y) - f(0, 0)| = |f(x, y)| \leq |x| \rightarrow 0 \text{ as } (x, y) \rightarrow (0, 0).$$

6.4.2 The product rules

Proposition 6.42. *Let $(X, \|\cdot\|_X)$, $(Y, \|\cdot\|_Y)$ be normed vector spaces, $\mathcal{U} \subseteq X$ be open, and $f : \mathcal{U} \rightarrow Y$ and $g : \mathcal{U} \rightarrow \mathbb{F}$ be differentiable at $x_0 \in \mathcal{U}$, where \mathbb{F} is the scalar field associated with the vector space Y . Then $gf : \mathcal{U} \rightarrow Y$ is differentiable at x_0 , and*

$$D(gf)(x_0)(v) = g(x_0)(Df)(x_0)(v) + (Dg)(x_0)(v)f(x_0). \quad (6.4.2)$$

Moreover, if $g(x_0) \neq 0$, then $\frac{f}{g} : \mathcal{U} \rightarrow Y$ is also differentiable at x_0 , and $D(\frac{f}{g})(x_0) : X \rightarrow Y$ is given by

$$D\left(\frac{f}{g}\right)(x_0)(v) = \frac{g(x_0)((Df)(x_0)(v)) - (Dg)(x_0)(v)f(x_0)}{g^2(x_0)}. \quad (6.4.3)$$

Proof. We only prove (6.4.2), and (6.4.3) is left as an exercise.

Define $Av = g(x_0)(Df)(x_0)(v) + (Dg)(x_0)(v)f(x_0)$. Then $A \in \mathcal{B}(X, Y)$. Moreover,

$$\begin{aligned} (gf)(x) - (gf)(x_0) - A(x - x_0) &= g(x_0)(f(x) - f(x_0) - (Df)(x_0)(x - x_0)) \\ &\quad + (g(x) - g(x_0) - (Dg)(x_0)(x - x_0))f(x_0) \\ &\quad + ((Dg)(x_0)(x - x_0))(f(x) - f(x_0)). \end{aligned}$$

Since $(Dg)(x_0) \in \mathcal{B}(X, \mathbb{F})$, $\|(Dg)(x_0)\|_{\mathcal{B}(X, \mathbb{F})} < \infty$; thus using the inequality

$$|(Dg)(x_0)(x - x_0)| \leq \|(Dg)(x_0)\|_{\mathcal{B}(X, \mathbb{F})} \|x - x_0\|_X$$

and the continuity of f at x_0 (due to Theorem 6.37), we find that

$$\lim_{x \rightarrow x_0} \left| \frac{|(Dg)(x_0)(x - x_0)|}{\|x - x_0\|_X} \|f(x) - f(x_0)\|_Y \right| \leq \lim_{x \rightarrow x_0} \|(Dg)(x_0)\|_{\mathcal{B}(X, \mathbb{F})} \|f(x) - f(x_0)\|_Y = 0.$$

As a consequence,

$$\begin{aligned} &\lim_{x \rightarrow x_0} \frac{\|(gf)(x) - (gf)(x_0) - A(x - x_0)\|_Y}{\|x - x_0\|_X} \\ &\leq |g(x_0)| \lim_{x \rightarrow x_0} \frac{\|f(x) - f(x_0) - (Df)(x_0)(x - x_0)\|_Y}{\|x - x_0\|_X} \\ &\quad + \lim_{x \rightarrow x_0} \left[\frac{|g(x) - g(x_0) - (Dg)(x_0)(x - x_0)|}{\|x - x_0\|_X} \|f(x)\|_Y \right] \\ &\quad + \lim_{x \rightarrow x_0} \left[\frac{|(Dg)(x_0)(x - x_0)|}{\|x - x_0\|_X} \|f(x) - f(x_0)\|_Y \right] = 0 \end{aligned}$$

which implies that gf is differentiable at x_0 with derivative $D(gf)(x_0)$ given by (6.4.2). \square

6.4.3 The chain rule

Theorem 6.43. Let $(X, \|\cdot\|_X)$, $(Y, \|\cdot\|_Y)$, $(Z, \|\cdot\|_Z)$ be normed vector spaces, $\mathcal{U} \subseteq X$ and $\mathcal{V} \subseteq Y$ be open sets. Suppose that $f : \mathcal{U} \rightarrow Y$ is differentiable at $x_0 \in \mathcal{U}$, $f(\mathcal{U}) \subseteq \mathcal{V}$, and $g : \mathcal{V} \rightarrow Z$ is differentiable at $f(x_0)$. Then the map $F = g \circ f : \mathcal{U} \rightarrow Z$ defined by

$$F(x) = g(f(x)) \quad \forall x \in \mathcal{U}$$

is differentiable at x_0 , and

$$(DF)(x_0)(h) = (Dg)(f(x_0))((Df)(x_0)(h)) \quad \forall h \in X.$$

In particular, if $X = \mathbb{R}^n$, $Y = \mathbb{R}^m$ and $Z = \mathbb{R}^\ell$, then

$$((DF)(x_0))_{ij} = \sum_{k=1}^m \frac{\partial g_i}{\partial y_k}(f(x_0)) \frac{\partial f_k}{\partial x_j}(x_0).$$

Proof. To simplify the notation, let $y_0 = f(x_0)$, $A = (Df)(x_0) \in \mathcal{B}(X, Y)$, and $B = (Dg)(y_0) \in \mathcal{B}(Y, Z)$. Let $\varepsilon > 0$ be given. By the differentiability of f and g at x_0 and y_0 , there exists $\delta_1, \delta_2 > 0$ such that if $\|x - x_0\|_X < \delta_1$ and $\|y - y_0\|_Y < \delta_2$, we have

$$\begin{aligned} \|f(x) - f(x_0) - A(x - x_0)\|_Y &\leq \min\left\{1, \frac{\varepsilon}{2(\|B\| + 1)}\right\} \|x - x_0\|_X, \\ \|g(y) - g(y_0) - B(y - y_0)\|_Z &\leq \frac{\varepsilon}{2(\|A\| + 1)} \|y - y_0\|_Y. \end{aligned}$$

Define

$$\begin{aligned} u(h) &= f(x_0 + h) - f(x_0) - Ah & \forall \|h\|_X < \delta_1, \\ v(k) &= g(y_0 + k) - g(y_0) - Bk & \forall \|k\|_Y < \delta_2. \end{aligned}$$

Then if $\|h\|_X < \delta_1$ and $\|k\|_Y < \delta_2$,

$$\|u(h)\|_Y \leq \|h\|_X, \quad \|u(h)\|_Y \leq \frac{\varepsilon}{2(\|B\| + 1)} \|h\|_X \quad \text{and} \quad \|v(k)\|_Z \leq \frac{\varepsilon}{2(\|A\| + 1)} \|k\|_Y.$$

Let $k = f(x_0 + h) - f(x_0) = Ah + u(h)$. Then $\lim_{h \rightarrow 0} k = 0$; thus there exists $\delta_3 > 0$ such that

$$\|k\|_Y < \delta_2 \quad \text{whenever} \quad \|h\|_X < \delta_3.$$

Since

$$\begin{aligned} F(x_0 + h) - F(x_0) &= g(y_0 + k) - g(y_0) = Bk + v(k) = B(Ah + u(h)) + v(k) \\ &= BAh + Bu(h) + v(k), \end{aligned}$$

we conclude that if $\|h\|_X < \delta = \min\{\delta_1, \delta_3\}$,

$$\begin{aligned} \|F(x_0 + h) - F(x_0) - BAh\|_Z &\leq \|Bu(h)\|_Z + \|v(k)\|_Z \leq \|B\| \|u(h)\|_Y + \frac{\varepsilon}{2(\|A\| + 1)} \|k\|_Y \\ &\leq \frac{\varepsilon}{2} \|h\|_X + \frac{\varepsilon}{2(\|A\| + 1)} (\|A\| \|h\|_X + \|u(h)\|_Y) \leq \frac{\varepsilon}{2} \|h\|_X + \frac{\varepsilon}{2} \|h\|_X = \varepsilon \|h\|_X \end{aligned}$$

which implies that F is differentiable at x_0 and $(DF)(x_0) = BA$. \square

Example 6.44. Consider the polar coordinate $x = r \cos \theta$, $y = r \sin \theta$. Then every function $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ is associated with a function $F : [0, \infty) \times [0, 2\pi) \rightarrow \mathbb{R}$ satisfying

$$F(r, \theta) = f(r \cos \theta, r \sin \theta).$$

Suppose that f is differentiable. Then F is differentiable, and the chain rule implies that

$$\begin{bmatrix} \frac{\partial F}{\partial r} & \frac{\partial F}{\partial \theta} \end{bmatrix} = \begin{bmatrix} \frac{\partial f}{\partial x} & \frac{\partial f}{\partial y} \end{bmatrix} \begin{bmatrix} \frac{\partial x}{\partial r} & \frac{\partial x}{\partial \theta} \\ \frac{\partial y}{\partial r} & \frac{\partial y}{\partial \theta} \end{bmatrix} = \begin{bmatrix} \frac{\partial f}{\partial x} & \frac{\partial f}{\partial y} \end{bmatrix} \begin{bmatrix} \cos \theta & -r \sin \theta \\ \sin \theta & r \cos \theta \end{bmatrix}.$$

Therefore, we arrive at the following form of chain rule

$$\frac{\partial}{\partial r} = \frac{\partial x}{\partial r} \frac{\partial}{\partial x} + \frac{\partial y}{\partial r} \frac{\partial}{\partial y} \quad \text{and} \quad \frac{\partial}{\partial \theta} = \frac{\partial x}{\partial \theta} \frac{\partial}{\partial x} + \frac{\partial y}{\partial \theta} \frac{\partial}{\partial y}$$

which is commonly seen in Calculus textbook.

Example 6.45. Let $f : \mathbb{R} \rightarrow \mathbb{R}$ and $F : \mathbb{R}^2 \rightarrow \mathbb{R}$ be differentiable, and $F(x, f(x)) = 0$ and $\frac{\partial F}{\partial y} \neq 0$. Then $f'(x) = -\frac{F_x(x, f(x))}{F_y(x, f(x))}$, where $F_x = \frac{\partial F}{\partial x}$ and $F_y = \frac{\partial F}{\partial y}$.

Example 6.46. Let $\gamma : (0, 1) \rightarrow \mathbb{R}^n$ and $f : \mathbb{R}^n \rightarrow \mathbb{R}$ be differentiable. Let $F(t) = f(\gamma(t))$. Then $F'(t) = (Df)(\gamma(t))\gamma'(t)$.

Example 6.47. Let $f(u, v, w) = u^2v + vw^2$ and $g(x, y) = (xy, \sin x, e^x)$. Let $h = f \circ g : \mathbb{R}^2 \rightarrow \mathbb{R}$. Find $\frac{\partial h}{\partial x}$.

Way I: Compute $\frac{\partial h}{\partial x}$ directly: Since

$$h(x, y) = f(g(x, y)) = f(xy, \sin x, e^x) = x^2y^2 \sin x + e^x \sin^2 x,$$

we have

$$\frac{\partial h}{\partial x} = 2xy^2 \sin x + x^2y^2 \cos x + e^x \sin^2 x + 2e^x \cos x.$$

Way II: Use the chain rule:

$$\begin{aligned}\frac{\partial h}{\partial x} &= \frac{\partial f}{\partial u} \frac{\partial g_1}{\partial x} + \frac{\partial f}{\partial v} \frac{\partial g_2}{\partial x} + \frac{\partial f}{\partial w} \frac{\partial g_3}{\partial x} = 2uv \cdot y + (u^2 + 2wv) \cdot \cos x + v^2 \cdot e^x \\ &= 2xy^2 \sin x + (x^2y^2 + 2e^x \sin x) \cos x + e^x \sin^2 x.\end{aligned}$$

Example 6.48. Let $F(x, y) = f(x^2 + y^2)$, $f : \mathbb{R} \rightarrow \mathbb{R}$, $F : \mathbb{R}^2 \rightarrow \mathbb{R}$. Show that $x \frac{\partial F}{\partial y} = y \frac{\partial F}{\partial x}$.

Proof: Let $g(x, y) = x^2 + y^2$, $g : \mathbb{R}^2 \rightarrow \mathbb{R}$, then $F(x, y) = (f \circ g)(x, y)$. By the chain rule,

$$\begin{bmatrix} \frac{\partial F}{\partial x} & \frac{\partial F}{\partial y} \end{bmatrix} = f'(g(x, y)) \cdot \begin{bmatrix} \frac{\partial g}{\partial x} & \frac{\partial g}{\partial y} \end{bmatrix} = f'(g(x, y)) [2x \quad 2y]$$

which implies that

$$\frac{\partial F}{\partial x} = 2xf'(g(x, y)), \quad \frac{\partial F}{\partial y} = 2yf'(g(x, y)).$$

$$\text{So } y \frac{\partial F}{\partial x} = f'(g(x, y))2xy = x \frac{\partial F}{\partial y}.$$

6.4.4 The mean value theorem

Theorem 6.49. Let $\mathcal{U} \subseteq \mathbb{R}^n$ be open, and $f : \mathcal{U} \rightarrow \mathbb{R}^m$ with $f = (f_1, \dots, f_m)$. Suppose that f is differentiable on \mathcal{U} and the line segment joining x and y lies in \mathcal{U} . Then there exist points c_1, \dots, c_m on that segment such that

$$f_i(y) - f_i(x) = (Df_i)(c_i)(y - x) \quad \forall i = 1, \dots, m.$$

Moreover, if \mathcal{U} is convex and $\sup_{x \in \mathcal{U}} \|(Df)(x)\|_{\mathcal{B}(\mathbb{R}^n, \mathbb{R}^m)} \leq M$, then

$$\|f(x) - f(y)\|_{\mathbb{R}^m} \leq M \|x - y\|_{\mathbb{R}^n} \quad \forall x, y \in \mathcal{U}.$$

Proof. Let $\gamma : [0, 1] \rightarrow \mathbb{R}^n$ be given by $\gamma(t) = (1 - t)x + ty$. Then by Theorem 6.43, for each $i = 1, \dots, m$, $(f_i \circ \gamma) : [0, 1] \rightarrow \mathbb{R}$ is differentiable on $(0, 1)$; thus the mean value theorem (Corollary 4.65) implies that there exists $t_i \in (0, 1)$ such that

$$f_i(y) - f_i(x) = (f_i \circ \gamma)(1) - (f_i \circ \gamma)(0) = (f_i \circ \gamma)'(t_i) = (Df_i)(c_i)(\gamma'(t_i)),$$

where $c_i = \gamma(t_i)$. On the other hand, $\gamma'(t_i) = y - x$.

Let $g(t) = (f \circ \gamma)(t)$. Then the chain rule implies that $g'(t) = (Df)(\gamma(t))(y - x)$; thus

$$\|g'(t)\|_{\mathbb{R}^m} \leq \|(Df)(\gamma(t))\|_{\mathcal{B}(\mathbb{R}^n, \mathbb{R}^m)} \|y - x\|_{\mathbb{R}^n} \leq M \|y - x\|_{\mathbb{R}^n}.$$

Define $h(t) = (g(1) - g(0)) \cdot g(t)$. Then $h : [0, 1] \rightarrow \mathbb{R}$ is differentiable; thus by the mean value theorem (Corollary 4.65) we find that there exists $\xi \in (0, 1)$ such that

$$h(1) - h(0) = h'(\xi) = (g(1) - g(0)) \cdot g'(\xi);$$

thus by the fact that $g(0) = f(x)$ and $g(1) = f(y)$,

$$\begin{aligned} \|f(x) - f(y)\|_{\mathbb{R}^m}^2 &= h(1) - h(0) \leq \|g(1) - g(0)\|_{\mathbb{R}^m} \|g'(\xi)\|_{\mathbb{R}^m} \\ &\leq M \|f(x) - f(y)\|_{\mathbb{R}^m} \|x - y\|_{\mathbb{R}^n} \end{aligned}$$

which concludes the theorem. \square

Example 6.50. Let $f : [0, 1] \rightarrow \mathbb{R}^2$ be given by $f(t) = (t^2, t^3)$. Then there is no $s \in (0, 1)$ such that

$$(1, 1) = f(1) - f(0) = f'(s)(1 - 0) = f'(s)$$

since $f'(s) = (2s, 3s^2) \neq (1, 1)$ for all $s \in (0, 1)$.

Example 6.51. Let $f : \mathbb{R} \rightarrow \mathbb{R}^2$ be given by $f(x) = (\cos x, \sin x)$. Then $f(2\pi) - f(0) = (0, 0)$; however, $f'(x) = (-\sin x, \cos x)$ which cannot be a zero vector.

Example 6.52. Let f be given in Example 6.30, and \mathcal{U} be a small neighborhood of the curve

$$\mathcal{C} = \{(x, y) \mid x^2 + y^2 = 1, x \leq 0\} \cup \{(x, \pm 1) \mid 0 \leq x \leq 1\}.$$

Then

$$f(1, -1) - f(1, 1) = \frac{3\pi}{2}.$$

On the other hand,

$$(Df)(x, y)(0, -2) = \begin{bmatrix} \frac{-y}{x^2 + y^2} & \frac{x}{x^2 + y^2} \end{bmatrix} \begin{bmatrix} 0 \\ -2 \end{bmatrix} = -\frac{2x}{x^2 + y^2}$$

which can never be $\frac{3\pi}{2}$ since $\left| \frac{2x}{x^2 + y^2} \right| \leq 3$ if $(x, y) \in \mathcal{U}$ while $\frac{3\pi}{2} > 3$. Therefore, no point (x, y) in \mathcal{U} validates

$$(Df)(x, y)((1, -1) - (1, 1)) = f(1, -1) - f(1, 1).$$

Example 6.53. Suppose that $A \subseteq \mathbb{R}^n$ is an open convex set, and $f : A \rightarrow \mathbb{R}^m$ is differentiable and $Df(x) = 0$ for all $x \in A$. Then f is a constant; that is, for some $\alpha \in \mathbb{R}^m$ we have $f(x) = \alpha$ for all $x \in A$.

Reason: Since A is convex, then the Mean Value Theorem can be applied to any $x, y \in A$ such that $f_i(x) - f_i(y) = Df_i(c_i)(x - y) = 0$ ($\because Df_i = 0$) for $i = 1, 2, \dots, m$; thus $f(x) = f(y)$ for any $x, y \in A$. Let $\alpha = f(x) \in \mathbb{R}^m$, then we reach the conclusion.

Example 6.54. Let $f : [0, \infty) \rightarrow \mathbb{R}$ be continuous and be differentiable on $(0, \infty)$. Suppose that $f(0) = 0$ and $f'(x)$ is non-decreasing (that is, if $x < y$, then $f'(x) \leq f'(y)$). Show that $g : (0, \infty) \rightarrow \mathbb{R}$, $g(x) = \frac{f(x)}{x}$ is also non-decreasing.

Proof: It suffices to prove $g'(x) \geq 0$. Since f is differentiable on $(0, \infty)$, then g is differentiable on $(0, \infty)$, and $g'(x) = \frac{xf'(x) - f(x)}{x^2}$. Hence

$$g'(x) \geq 0 \Leftrightarrow xf'(x) \geq f(x).$$

Let $x > 0$ be fixed. Applying the Mean Value Theorem to f we find that

$$\exists c \in (0, x) \ni f(x) - f(0) = f'(c)(x - 0) \leq xf'(x).$$

6.5 Directional Derivatives and Gradient Vectors

Definition 6.55. Let f be real-valued and defined on a neighborhood of $x_0 \in \mathbb{R}^n$, and let $v \in \mathbb{R}^n$ be a unit vector. Then

$$(D_v f)(x_0) \equiv \left. \frac{d}{dt} \right|_{t=0} f(x_0 + tv) = \lim_{t \rightarrow 0} \frac{f(x_0 + tv) - f(x_0)}{t}$$

is called the **directional derivative** (方向導數) of f at x_0 in the direction v .

Remark 6.56. Let $\{e_j\}_{j=1}^n$ be the standard basis of \mathbb{R}^n . Then the partial derivative $\frac{\partial f}{\partial x_j}(x_0)$ (if it exists) is the directional derivative of f at x_0 in the direction e_j .

Theorem 6.57. Let $\mathcal{U} \subseteq \mathbb{R}^n$ be open, and $f : \mathcal{U} \rightarrow \mathbb{R}$ be differentiable at x_0 . Then the directional derivative of f at x_0 in the direction v is $(Df)(x_0)(v)$.

Proof. Let $\varepsilon > 0$ be given. Since f is differentiable at x_0 , there exists $\delta > 0$ such that

$$|f(x) - f(x_0) - (Df)(x_0)(x - x_0)| \leq \frac{\varepsilon}{2} \|x - x_0\|_{\mathbb{R}^n} \text{ whenever } \|x - x_0\|_{\mathbb{R}^n} < \delta.$$

In particular, if $x = x_0 + tv$ with v being a unit vector in \mathbb{R}^n and $0 < |t| < \delta$, then

$$\begin{aligned} \left| \frac{f(x_0 + tv) - f(x_0)}{t} - (Df)(x_0)(v) \right| &= \frac{|f(x_0 + tv) - f(x_0) - (Df)(x_0)(tv)|}{|t|} \\ &= \frac{|f(x) - f(x_0) - (Df)(x_0)(x - x_0)|}{|t|} \leq \frac{\varepsilon}{2} < \varepsilon; \end{aligned}$$

thus $(D_v f)(x_0) = (Df)(x_0)(v)$. □

Remark 6.58. When $v \in \mathbb{R}^n$ but $0 < \|v\|_{\mathbb{R}^n} \neq 1$, we let $\tilde{v} = \frac{v}{\|v\|_{\mathbb{R}^n}}$. Then the direction derivatives of a function $f : \mathcal{U} \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$ at $a \in \mathcal{U}$ in the direction v is

$$(D_v f)(a) = \lim_{t \rightarrow 0} \frac{f(a + tv) - f(a)}{t}.$$

Making a change of variable $s = \frac{t}{\|v\|_{\mathbb{R}^n}}$. Then

$$(Df)(x_0)(v) = \|v\|_{\mathbb{R}^n} (Df)(x_0)(\tilde{v}) = \|v\|_{\mathbb{R}^n} \lim_{t \rightarrow 0} \frac{f(a + tv) - f(a)}{t} = \lim_{s \rightarrow 0} \frac{f(a + sv) - f(a)}{s}.$$

We sometimes also call the value $(Df)(x_0)(v)$ the “directional derivative” of f in the “direction” v .

Example 6.59. The existence of directional derivatives of a function f at x_0 in all directions does not guarantee the differentiability of f at x_0 . For example, let $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ be given as in Example 6.41, and $v = (v_1, v_2) \in \mathbb{R}^2$ be a unit vector. Then

$$(D_v f)(0) = \lim_{t \rightarrow 0} \frac{f(tv_1, tv_2) - f(0, 0)}{t} = v_1^3.$$

However, f is not differentiable at $(0, 0)$. We also note that in this example, $(D_v f)(0) \neq (Jf)(0)v$, where $(Jf)(0) = \begin{bmatrix} \frac{\partial f}{\partial x}(0, 0) & \frac{\partial f}{\partial y}(0, 0) \end{bmatrix} = [1 \ 0]$ is the Jacobian matrix of f at $(0, 0)$.

Example 6.60. The existence of directional derivatives of a function f at x_0 in all directions does not even guarantee the continuity of f at x_0 . For example, let $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ be given by

$$f(x, y) = \begin{cases} \frac{xy^2}{x^2 + y^4} & \text{if } (x, y) \neq (0, 0), \\ 0 & \text{if } (x, y) = (0, 0), \end{cases}$$

and $v = (v_1, v_2) \in \mathbb{R}^2$ be a unit vector. Then if $v_1 \neq 0$,

$$(D_v f)(0) = \lim_{t \rightarrow 0} \frac{f(tv_1, tv_2) - f(0, 0)}{t} = \lim_{t \rightarrow 0} \frac{t^3 v_1 v_2^2}{t(t^2 v_1^2 + t^4 v_2^4)} = \frac{v_2^2}{v_1}$$

while if $v_1 = 0$,

$$(D_v f)(0) = \lim_{t \rightarrow 0} \frac{f(tv_1, tv_2) - f(0, 0)}{t} = 0.$$

However, f is not continuous at $(0, 0)$ since if (x, y) approaches $(0, 0)$ along the curve $x = my^2$ with $m \neq 0$, we have

$$\lim_{y \rightarrow 0} f(my^2, y) = \lim_{y \rightarrow 0} \frac{my^4}{m^2y^4 + y^4} = \frac{m}{m^2 + 1}$$

which depends on m . Therefore, f is not continuous at $(0, 0)$.

Example 6.61. Here comes another example showing that a function having directional derivative in all directions might not be continuous. Let $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ be given by

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

and $v = (v_1, v_2) \in \mathbb{R}^2$ be a unit vector. Then if $v_1 \neq 0$,

$$(D_v f)(0) = \lim_{t \rightarrow 0} \frac{f(tv_1, tv_2) - f(0, 0)}{t} = \lim_{t \rightarrow 0} \frac{t^2 v_1 v_2}{t(tv_1 + t^2 v_2^2)} = v_2$$

while if $v_1 = 0$,

$$(D_v f)(0) = \lim_{t \rightarrow 0} \frac{f(tv_1, tv_2) - f(0, 0)}{t} = 0.$$

However, f is not continuous at $(0, 0)$ since if (x, y) approaches $(0, 0)$ along the polar curve

$$\theta(r) = \frac{\pi}{2} + \sin^{-1}(r - mr^2) \quad 0 < r \ll 1,$$

we have

$$\begin{aligned} \lim_{\substack{(x,y) \rightarrow (0,0) \\ x=r \cos \theta(r), y=r \sin \theta(r)}} f(x, y) &= \lim_{r \rightarrow 0^+} \frac{r^2 \cos \theta(r) \sin \theta(r)}{r^2 \sin^2 \theta(r) + r \cos \theta(r)} = \lim_{r \rightarrow 0^+} \frac{r(-r + mr^2) \sin \theta(r)}{r \sin^2 \theta(r) - r + mr^2} \\ &= \lim_{r \rightarrow 0^+} \frac{(-r + mr^2) \sin \theta(r)}{\sin^2 \theta(r) - 1 + mr} = \frac{-1}{m} \end{aligned}$$

which depends on m . Therefore, f is not continuous at $(0, 0)$.

Definition 6.62. Let $\mathcal{U} \subseteq \mathbb{R}^n$ be an open set. The derivative of a scalar function $f : \mathcal{U} \rightarrow \mathbb{R}$ is called the **gradient** of f and is denoted by $\text{grad} f$ or ∇f .

Let $\mathcal{U} \subseteq \mathbb{R}^n$ be an open set, $a \in \mathcal{U}$ and $f : \mathcal{U} \rightarrow \mathbb{R}$ be a real-valued function. Suppose that $f \in \mathcal{C}^1(\mathcal{U}; \mathbb{R})$ and $(\nabla f)(a) \neq 0$. Then $\frac{\partial f}{\partial x_k}(a) \neq 0$ for some $1 \leq k \leq n$. W.L.O.G., we can assume that $\frac{\partial f}{\partial x_n}(a) \neq 0$. By the implicit function theorem, there exists an open neighborhood $\mathcal{V} \subseteq \mathbb{R}^{n-1}$ of (a_1, \dots, a_{n-1}) and an open neighborhood $\mathcal{W} \subseteq \mathbb{R}$ of a_n , as well as a \mathcal{C}^1 -function $\varphi : \mathcal{V} \rightarrow \mathbb{R}$ such that in a neighborhood of a the level set $\{x \in \mathcal{U} \mid f(x) = f(a)\}$ can be represented by $x_n = \varphi(x_1, \dots, x_{n-1})$; that is,

$$f(x_1, \dots, x_{n-1}, \varphi(x_1, \dots, x_{n-1})) = f(a) \quad \forall (x_1, \dots, x_{n-1}) \in \mathcal{V}.$$

Moreover,

$$\varphi_{x_j}(x_1, \dots, x_{n-1}) = -\frac{f_{x_j}(x_1, \dots, x_{n-1}, \varphi(x_1, \dots, x_{n-1}))}{f_{x_n}(x_1, \dots, x_{n-1}, \varphi(x_1, \dots, x_{n-1}))} \quad \forall (x_1, \dots, x_{n-1}) \in \mathcal{V}.$$

Consider the collection of vectors $\{v_j\}_{j=1}^{n-1}$ given by

$$v_j = \frac{\partial}{\partial x_j} \Big|_{x=a} (x_1, \dots, x_{n-1}, \varphi(x_1, \dots, x_{n-1})) \quad (x_1, \dots, x_{n-1}) \in \mathcal{V}.$$

Then v_j 's are tangent vectors of the level surface. If $\{e_j\}_{j=1}^n$ is the standard basis of \mathbb{R}^n , then

$$v_j = e_j + (0, \dots, 0, \varphi_{x_j}(a_1, \dots, a_{n-1})) = e_j - \left(0, \dots, 0, \frac{f_{x_j}(a)}{f_{x_n}(a)}\right).$$

Therefore, the gradient vector $(\nabla f)(a)$ is perpendicular to v_j for all $1 \leq j \leq n-1$ which conclude the following

Proposition 6.63. *Let $\mathcal{U} \subseteq \mathbb{R}^n$ be open and $f \in \mathcal{C}^1(\mathcal{U}; \mathbb{R})$; that is, $f : \mathcal{U} \rightarrow \mathbb{R}$ is continuously differentiable. Then if $(\nabla f)(x_0) \neq 0$, the vector $\frac{(\nabla f)(x_0)}{\|(\nabla f)(x_0)\|_{\mathbb{R}^n}}$ is the unit normal to the level set $\{x \in \mathcal{U} \mid f(x) = f(x_0)\}$ at x_0 .*

Example 6.64. Find the normal to $\mathcal{S} = \{(x, y, z) \mid x^2 + y^2 + z^2 = 3\}$ at $(1, 1, 1) \in \mathcal{S}$.

Solution: Take $f(x, y, z) = x^2 + y^2 + z^2 - 3$. Then $(\nabla f)(x, y, z) = (2x, 2y, 2z)$; thus $(\nabla f)(1, 1, 1) = (2, 2, 2)$ is normal to \mathcal{S} at $(1, 1, 1)$.

Example 6.65. Consider the surface

$$\mathcal{S} = \{(x, y, z) \in \mathbb{R}^3 \mid x^2 - y^2 + xyz = 1\}.$$

Find the tangent plane of \mathcal{S} at $(1, 0, 1)$.

Solution: Let $f(x, y, z) = x^2 - y^2 + xyz$. Then

$$\mathcal{S} = \{(x, y, z) \in \mathbb{R}^3 \mid f(x, y, z) = f(1, 0, 1)\};$$

that is, \mathcal{S} is a level set of f . Since $(\nabla f)(1, 0, 1) = (2, 1, 0) \neq (0, 0, 0)$, $(2, 1, 0)$ is normal to \mathcal{S} at $(1, 0, 1)$; thus the tangent plane of \mathcal{S} at $(1, 0, 1)$ is $2(x - 1) + y = 0$. \square

Proposition 6.66. *Let $f : \mathbb{R}^n \rightarrow \mathbb{R}$ be differentiable. Then $\pm \frac{\nabla f}{\|\nabla f\|_{\mathbb{R}^n}}$ is the direction in which the function f increases/decreases most rapidly (最速上升/下降方向).*

Proof. Let $x_0 \in \mathbb{R}^n$ be given. Suppose that f increases most rapidly in the direction v , then $(D_v f)(x_0) = \sup_{\|w\|_{\mathbb{R}^n}=1} (D_w f)(x_0)$. Since f is differentiable, $(D_w f)(x_0) = (Df)(x_0)(w) = (\nabla f)(x_0) \cdot w$ which is maximized in the direction $\frac{(\nabla f)(x_0)}{\|(\nabla f)(x_0)\|_{\mathbb{R}^n}}$. \square

Example 6.67. Let $f : \mathbb{R}^3 \rightarrow \mathbb{R}$ be given by $f(x, y, z) = x^2 y \sin z$. Find the direction of the greatest rate of change at $(3, 2, 0)$.

Solution: We compute the gradient of f at $(3, 2, 0)$ as follows:

$$\begin{aligned} (\nabla f)(3, 2, 0) &= \left(\frac{\partial f}{\partial x}(3, 2, 0), \frac{\partial f}{\partial y}(3, 2, 0), \frac{\partial f}{\partial z}(3, 2, 0) \right) \\ &= (2xy \sin z, x^2 \sin z, x^2 y \cos z) \Big|_{(x,y,z)=(3,2,0)} = (0, 0, 18). \end{aligned}$$

Therefore, the direction of the greatest rate of change of f at $(3, 2, 0)$ is $(0, 0, 1)$.

6.6 Higher Order Derivatives of Functions

Let $\mathcal{U} \subseteq \mathbb{R}^n$ be open, and $f : \mathcal{U} \rightarrow \mathbb{R}^m$ is differentiable. By Proposition 6.5, the space $(\mathcal{B}(\mathbb{R}^n, \mathbb{R}^m), \|\cdot\|_{\mathcal{B}(\mathbb{R}^n, \mathbb{R}^m)})$ is a normed space (in fact, it is a Banach space), so it is legitimate to ask if $Df : \mathcal{U} \rightarrow \mathcal{B}(\mathbb{R}^n, \mathbb{R}^m)$ is differentiable or not. If Df is differentiable at x_0 , we call f twice differentiable at x_0 , and denote the twice derivative of f at x_0 as $(D^2 f)(x_0)$. If Df is differentiable on \mathcal{U} , then $D^2 f : \mathcal{U} \rightarrow \mathcal{B}(\mathbb{R}^n, \mathcal{B}(\mathbb{R}^n, \mathbb{R}^m))$. Similar, we can talk about three times differentiability of a function if it is twice differentiable. In general, we have the following

Definition 6.68. Let $(X, \|\cdot\|_X)$ and $(Y, \|\cdot\|_Y)$ be normed spaces, and $\mathcal{U} \subseteq X$ be open. A function $f : \mathcal{U} \rightarrow Y$ is said to be *twice differentiable* at $a \in \mathcal{U}$ if

1. f is (once) differentiable in a neighborhood of a ;
2. there exists $L_2 \in \mathcal{B}(X, \mathcal{B}(X, Y))$, usually denoted by $(D^2f)(a)$ and called the **second derivative** of f at a , such that

$$\lim_{x \rightarrow a} \frac{\|(Df)(x) - (Df)(a) - L_2(x - a)\|_{\mathcal{B}(X, Y)}}{\|x - a\|_X} = 0.$$

For any two vectors $u, v \in X$, $(D^2f)(a)(v) \in \mathcal{B}(X, Y)$ and $(D^2f)(a)(v)(u) \in Y$. The vector $(D^2f)(a)(v)(u)$ is usually denoted by $(D^2f)(a)(u, v)$.

In general, a function f is said to be **k -times differentiable** at $a \in \mathcal{U}$ if

1. f is $(k - 1)$ -times differentiable in a neighborhood of a ;
2. there exists $L_k \in \mathcal{B}(X, \underbrace{\mathcal{B}(X, \dots, \mathcal{B}(X, Y) \dots)}_{\substack{k \text{ copies of "X" } \\ k \text{ copies of "}"}})$, usually denoted by $(D^k f)(a)$ and called the **k -th derivative** of f at a , such that

$$\lim_{x \rightarrow a} \frac{\|(D^{k-1}f)(x) - (D^{k-1}f)(a) - L_k(x - a)\|_{\mathcal{B}(X, \mathcal{B}(X, \dots, \mathcal{B}(X, Y) \dots))}}{\|x - a\|_X} = 0.$$

For any k vectors $u^{(1)}, \dots, u^{(k)} \in X$, the vector $(D^k f)(a)(u^{(1)}, \dots, u^{(k)})$ is defined as the vector

$$(D^k f)(a)(u^{(k)})(u^{(k-1)}) \dots (u^{(1)}).$$

Example 6.69. Let $(X, \|\cdot\|_X)$ and $(Y, \|\cdot\|_Y)$ be two normed spaces, and $f(x) = Lx$ for some $L \in \mathcal{B}(X, Y)$. From Example 6.15, $(Df)(x_0) = L$ for all $x_0 \in X$; thus $(D^2f)(x_0) = 0$ since $Df : \mathcal{U} \in \mathcal{B}(X, Y)$ is a “constant” map. In fact, one can also conclude from

$$\lim_{x \rightarrow x_0} \frac{\|(Df)(x) - (Df)(x_0) - 0(x - x_0)\|_{\mathcal{B}(X, Y)}}{\|x - x_0\|_X} = 0$$

that $(D^2f)(x_0) = 0$ for all $x_0 \in X$.

Remark 6.70. We focus on what $(D^k f)(a)(u_k)(\dots)(u_1)$ means in this remark. We first look at the case that f is twice differentiable at a . With $x = a + tv$ for $v \in X$ with $\|v\|_X = 1$ in the definition, we find that

$$\lim_{t \rightarrow 0} \frac{\|(Df)(a + tv) - (Df)(a) - t(D^2f)(a)(v)\|_{\mathcal{B}(X, Y)}}{|t|} = 0.$$

Since $(Df)(a + tv) - (Df)(a) - t(D^2f)(a)(v) \in \mathcal{B}(X, Y)$, for all $u \in X$ with $\|u\|_X = 1$ we have

$$\begin{aligned} & \lim_{t \rightarrow 0} \frac{\|(Df)(a + tv)(u) - (Df)(a)(u) - t(D^2f)(a)(v)(u)\|_Y}{|t|} \\ &= \lim_{t \rightarrow 0} \frac{\|[(Df)(a + tv) - (Df)(a) - t(D^2f)(a)(v)](u)\|_Y}{|t|} \\ &\leq \lim_{t \rightarrow 0} \frac{\|(Df)(a + tv) - (Df)(a) - t(D^2f)(a)(v)\|_{\mathcal{B}(X, Y)}}{|t|} = 0. \end{aligned}$$

On the other hand, by the definition of the direction derivative,

$$(Df)(a + tv)(u) - (Df)(a)(u) = \lim_{s \rightarrow 0} \left[\frac{f(a + tv + su) - f(a + tv)}{s} - \frac{f(a + su) - f(a)}{s} \right];$$

thus the limit above implies that

$$\begin{aligned} (D^2f)(a)(v)(u) &= \lim_{t \rightarrow 0} \lim_{s \rightarrow 0} \frac{f(a + tv + su) - f(a + tv) - f(a + su) + f(a)}{st} \\ &= \lim_{t \rightarrow 0} \frac{\lim_{s \rightarrow 0} \frac{f(a + tv + su) - f(a + tv)}{s} - \lim_{s \rightarrow 0} \frac{f(a + su) - f(a)}{s}}{t} \\ &= D_v(D_u f)(a). \end{aligned}$$

Therefore, $(D^2f)(a)(v)(u)$ is obtained by first differentiating f near a in the u -direction, then differentiating (Df) at a in the v -direction.

In general, $(D^k f)(a)(u_k) \cdots (u_1)$ is obtained by first differentiating f near a in the u_1 -direction, then differentiating (Df) near a in the u_2 -direction, and so on, and finally differentiating $(D^{k-1} f)$ at a in the u_k -direction.

Remark 6.71. Since $(D^2f)(a) \in \mathcal{B}(X, \mathcal{B}(X, Y))$, if $v_1, v_2 \in X$ and $c \in \mathbb{R}$, we have $(D^2f)(a)(cv_1 + v_2) = c(D^2f)(a)(v_1) + (D^2f)(a)(v_2)$ (treated as “vectors” in $\mathcal{B}(X, Y)$); thus

$$(D^2f)(a)(cv_1 + v_2)(u) = c(D^2f)(a)(v_1)(u) + (D^2f)(a)(v_2)(u) \quad \forall u, v_1, v_2 \in X.$$

On the other hand, since $(D^2f)(a)(v) \in \mathcal{B}(X, Y)$,

$$(D^2f)(a)(v)(cu_1 + u_2) = c(D^2f)(a)(v)(u_1) + (D^2f)(a)(v)(u_2) \quad \forall u_1, u_2, v \in X.$$

Therefore, $(D^2f)(a)(v)(u)$ is linear in both u and v variables. A map with such kind of property is called a **bilinear** map (meaning 2-linear). In particular, $(D^2f)(a) : X \times X \rightarrow Y$ is a bilinear map.

In general, the vector $(D^k f)(a)(u^{(1)}, \dots, u^{(k)})$ is linear in $u^{(1)}, \dots, u^{(k)}$; that is,

$$\begin{aligned} (D^k f)(a)(u^{(1)}, \dots, u^{(i-1)}, \alpha v + \beta w, u^{(i+1)}, \dots, u^{(k)}) \\ = \alpha(D^k f)(a)(u^{(1)}, \dots, u^{(i-1)}, v, u^{(i+1)}, \dots, u^{(k)}) \\ + \beta(D^k f)(a)(u^{(1)}, \dots, u^{(i-1)}, w, u^{(i+1)}, \dots, u^{(k)}) \end{aligned}$$

for all $v, w \in X$, $\alpha, \beta \in \mathbb{R}$, and $i = 1, \dots, n$. Such kind of map which is linear in each component when the other $k - 1$ components are fixed is called ***k-linear***.

Consider the case that X is finite dimensional with $\dim(X) = n$, $\{e_1, e_2, \dots, e_n\}$ is a basis of X , and $Y = \mathbb{R}$. Then $(D^2 f)(a) : X \times X \rightarrow Y$ is a bilinear form (here the term “form” means that $Y = \mathbb{R}$). A bilinear form $B : X \times X \rightarrow \mathbb{R}$ can be represented as follows: Let $a_{ij} = B(e_i, e_j) \in \mathbb{R}$ for $i, j = 1, 2, \dots, n$. Given $x, y \in \mathbb{R}^n$, write $u = \sum_{i=1}^n u_i e_i$ and $v = \sum_{j=1}^n v_j e_j$. Then by the bilinearity of B ,

$$B(u, v) = B\left(\sum_{i=1}^n u_i e_i, \sum_{j=1}^n v_j e_j\right) = \sum_{i,j=1}^n u_i v_j a_{ij} = \begin{bmatrix} u_1 & \cdots & u_n \end{bmatrix} \begin{bmatrix} a_{11} & \cdots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{n1} & \cdots & a_{nn} \end{bmatrix} \begin{bmatrix} v_1 \\ \vdots \\ v_n \end{bmatrix}.$$

Therefore, if $f : \mathcal{U} \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$ is twice differentiable at a , then the bilinear form $(D^2 f)(a)$ can be represented as

$$(D^2 f)(a)(u, v) = \begin{bmatrix} u_1 & \cdots & u_n \end{bmatrix} \begin{bmatrix} (D^2 f)(a)(e_1, e_1) & \cdots & (D^2 f)(a)(e_1, e_n) \\ \vdots & \ddots & \vdots \\ (D^2 f)(a)(e_n, e_1) & \cdots & (D^2 f)(a)(e_n, e_n) \end{bmatrix} \begin{bmatrix} v_1 \\ \vdots \\ v_n \end{bmatrix}.$$

The following proposition is an analogy of Proposition 6.27. The proof is similar to the one of Proposition 6.27, and is left as an exercise.

Proposition 6.72. *Let $\mathcal{U} \subseteq \mathbb{R}^n$ be open, $x_0 \in \mathcal{U}$, and $f = (f_1, \dots, f_m) : \mathcal{U} \rightarrow \mathbb{R}^m$. Then f is k -times differentiable at x_0 if and only if f_i is k -times differentiable at x_0 for all $i = 1, \dots, m$.*

Due to the proposition above, when talking about the higher-order differentiability of $f : \mathcal{U} \subseteq \mathbb{R}^n \rightarrow \mathbb{R}^m$ and a point $x_0 \in \mathcal{U}$, from now on we only focus on the case $m = 1$.

Example 6.73. In this example, we focus on what the second derivative $(D^2 f)(a)$ of a function f is, or in particular, what $(D^2 f)(a)(e_i, e_j)$ (which appears in the Remark 6.71) is, if $X = \mathbb{R}^2$.

Let $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ be differentiable, then

$$[(Df)(x, y)] = [f_x(x, y) \quad f_y(x, y)] = \left[\frac{\partial f}{\partial x}(x, y) \quad \frac{\partial f}{\partial y}(x, y) \right].$$

Suppose that f is twice differentiable at (a, b) , and let $L_2 = (D^2f)(a, b)$. Then

$$\lim_{(x, y) \rightarrow (a, b)} \frac{\| (Df)(x, y) - (Df)(a, b) - L_2((x - a, y - b)) \|_{\mathcal{B}(\mathbb{R}^2, \mathbb{R})}}{\sqrt{(x - a)^2 + (y - b)^2}} = 0$$

or equivalently,

$$\lim_{(x, y) \rightarrow (a, b)} \frac{\| [f_x(x, y) \quad f_y(x, y)] - [f_x(a, b) \quad f_y(a, b)] - [L_2((x - a, y - b))] \|_{\mathcal{B}(\mathbb{R}^2, \mathbb{R})}}{\sqrt{(x - a)^2 + (y - b)^2}} = 0,$$

where $[L_2((x - a, y - b))]$ denotes the matrix representation of the linear map $L_2((x - a, y - b)) \in \mathcal{B}(\mathbb{R}^2, \mathbb{R})$. In particular, we must have

$$\lim_{x \rightarrow a} \left\| \begin{bmatrix} \frac{f_x(x, b) - f_x(a, b)}{x - a} & \frac{f_y(x, b) - f_y(a, b)}{x - a} \end{bmatrix} - [L_2 e_1] \right\|_{\mathcal{B}(\mathbb{R}^2, \mathbb{R})} = 0$$

and

$$\lim_{y \rightarrow b} \left\| \begin{bmatrix} \frac{f_x(a, y) - f_x(a, b)}{y - b} & \frac{f_y(a, y) - f_y(a, b)}{y - b} \end{bmatrix} - [L_2 e_2] \right\|_{\mathcal{B}(\mathbb{R}^2, \mathbb{R})} = 0.$$

Using the notation of second partial derivatives, we find that

$$[L_2 e_1] = [f_{xx}(a, b) \quad f_{yx}(a, b)] \quad \text{and} \quad [L_2 e_2] = [f_{xy}(a, b) \quad f_{yy}(a, b)],$$

where $f_{xy} = (f_x)_y = \frac{\partial}{\partial y} \left(\frac{\partial f}{\partial x} \right)$ and $f_{yx} = (f_y)_x = \frac{\partial}{\partial x} \left(\frac{\partial f}{\partial y} \right)$. Therefore, if $v = v_1 e_1 + v_2 e_2$,

$$[L_2 v] = [L_2(v_1 e_1 + v_2 e_2)] = [v_1 f_{xx}(a, b) + v_2 f_{xy}(a, b) \quad v_1 f_{yx}(a, b) + v_2 f_{yy}(a, b)]. \quad (6.6.1)$$

Symbolically, we can write

$$[L_2] = \begin{bmatrix} [f_{xx}(a, b) \quad f_{yx}(a, b)] & [f_{xy}(a, b) \quad f_{yy}(a, b)] \end{bmatrix}$$

so that

$$\begin{aligned} [L_2(v_1 e_1 + v_2 e_2)] &= [L_2] \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = \begin{bmatrix} [f_{xx}(a, b) \quad f_{yx}(a, b)] & [f_{xy}(a, b) \quad f_{yy}(a, b)] \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} \\ &= v_1 [f_{xx}(a, b) \quad f_{yx}(a, b)] + v_2 [f_{xy}(a, b) \quad f_{yy}(a, b)]. \end{aligned}$$

For two vectors \mathbf{u} and \mathbf{v} , what does $(D^2f)(a, b)(\mathbf{v})(\mathbf{u})$ or $(D^2f)(a, b)(\mathbf{u}, \mathbf{v})$ mean? To see this, let $\mathbf{u} = u_1\mathbf{e}_1 + u_2\mathbf{e}_2$ and $\mathbf{v} = v_1\mathbf{e}_1 + v_2\mathbf{e}_2$. Then

$$\begin{aligned} [(D^2f)(a, b)(\mathbf{v})(\mathbf{u})] &= [(D^2f)(a, b)(\mathbf{v})][\mathbf{u}] = [L_2(v_1\mathbf{e}_1 + v_2\mathbf{e}_2)] \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} \\ &= v_1 \begin{bmatrix} f_{xx}(a, b) & f_{yx}(a, b) \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} + v_2 \begin{bmatrix} f_{xy}(a, b) & f_{yy}(a, b) \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} \\ &= \begin{bmatrix} v_1 & v_2 \end{bmatrix} \begin{bmatrix} f_{xx}(a, b) & f_{yx}(a, b) \\ f_{xy}(a, b) & f_{yy}(a, b) \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix}. \end{aligned}$$

Therefore, $(D^2f)(a, b)(\mathbf{e}_1, \mathbf{e}_1) = f_{xx}(a, b)$, $(D^2f)(a, b)(\mathbf{e}_1, \mathbf{e}_2) = f_{xy}(a, b)$, $(D^2f)(a, b)(\mathbf{e}_2, \mathbf{e}_1) = f_{yx}(a, b)$ and $(D^2f)(a, b)(\mathbf{e}_2, \mathbf{e}_2) = f_{yy}(a, b)$.

On the other hand, we can identify $\mathcal{B}(\mathbb{R}^2; \mathbb{R})$ as \mathbb{R}^2 (every 1×2 matrix is a “row” vector), and treat $g \equiv [Df]^\top : \mathbb{R}^2 \rightarrow \mathbb{R}^2$ as a vector-valued function. By Theorem 6.21 $(Dg)(a, b)$ can be represented as a 2×2 matrix given by

$$[(Dg)(a, b)] = \begin{bmatrix} f_{xx}(a, b) & f_{xy}(a, b) \\ f_{yx}(a, b) & f_{yy}(a, b) \end{bmatrix}.$$

We note that the representation above means

$$\lim_{(x, y) \rightarrow (a, b)} \frac{\left\| \begin{bmatrix} f_x(x, y) \\ f_y(x, y) \end{bmatrix} - \begin{bmatrix} f_x(a, b) \\ f_y(a, b) \end{bmatrix} - \begin{bmatrix} f_{xx}(a, b) & f_{xy}(a, b) \\ f_{yx}(a, b) & f_{yy}(a, b) \end{bmatrix} \begin{bmatrix} x - a \\ y - b \end{bmatrix} \right\|_{\mathbb{R}^2}}{\sqrt{(x - a)^2 + (y - b)^2}} = 0.$$

The equality above is equivalent to that

$$\lim_{(x, y) \rightarrow (a, b)} \frac{\left\| [(Df)(x, y)] - [(Df)(a, b)] - [x - a \quad y - b] \begin{bmatrix} f_{xx}(a, b) & f_{yx}(a, b) \\ f_{xy}(a, b) & f_{yy}(a, b) \end{bmatrix} \right\|_{\mathbb{R}^2}}{\sqrt{(x - a)^2 + (y - b)^2}} = 0$$

According to the equality above, $L_2 = (D^2f)(a, b)$ should be defined by

$$[L_2(v_1\mathbf{e}_1 + v_2\mathbf{e}_2)] = \begin{bmatrix} v_1 & v_2 \end{bmatrix} \begin{bmatrix} f_{xx}(a, b) & f_{yx}(a, b) \\ f_{xy}(a, b) & f_{yy}(a, b) \end{bmatrix} = \left(\begin{bmatrix} f_{xx}(a, b) & f_{xy}(a, b) \\ f_{yx}(a, b) & f_{yy}(a, b) \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} \right)^\top$$

which agrees with what (6.6.1) provides.

Proposition 6.74. *Let $\mathcal{U} \subseteq \mathbb{R}^n$ be open, and $f : \mathcal{U} \rightarrow \mathbb{R}$. Suppose that f is k -times differentiable at a . Then for k vectors $u^{(1)}, \dots, u^{(k)} \in \mathbb{R}^n$,*

$$\begin{aligned} (D^k f)(a)(u^{(1)}, \dots, u^{(k)}) &= \sum_{j_1, \dots, j_k=1}^n \frac{\partial^k f}{\partial x_{j_k} \partial x_{j_{k-1}} \cdots \partial x_{j_1}}(a) u_{j_1}^{(1)} u_{j_2}^{(2)} \cdots u_{j_k}^{(k)} \\ &= \sum_{j_1, \dots, j_k=1}^n \frac{\partial}{\partial x_{j_k}} \left(\frac{\partial}{\partial x_{j_{k-1}}} \left(\cdots \frac{\partial}{\partial x_{j_2}} \left(\frac{\partial f}{\partial x_{j_1}} \right) \cdots \right) \right) (a) u_{j_1}^{(1)} u_{j_2}^{(2)} \cdots u_{j_k}^{(k)}, \end{aligned}$$

where $u^{(i)} = (u_1^{(i)}, u_2^{(i)}, \dots, u_n^{(i)})$ for all $i = 1, \dots, k$. (上標括號中的數字指所給定的 k 個向量中的第幾個向量，下標指每一個固定向量的第幾個分量)

Proof. Let $\{e_j\}_{j=1}^n$ be the standard basis of \mathbb{R}^n . By Remark 6.71 (on multi-linearity), it suffices to show that

$$(D^k f)(a)(e_{j_k})(e_{j_{k-1}}) \cdots (e_{j_2})(e_{j_1}) = (D^k f)(a)(e_{j_1}, \dots, e_{j_k}) = \frac{\partial^k f}{\partial x_{j_k} \partial x_{j_{k-1}} \cdots \partial x_{j_1}}(a) \quad (6.6.2)$$

provided that f is k -times differentiable at a since if so, we must have

$$\begin{aligned} (D^k f)(a)(u^{(1)}, \dots, u^{(k)}) &= (D^k f)(a) \left(\sum_{j_1=1}^n u_{j_1}^{(1)} e_{j_1}, \dots, \sum_{j_k=1}^n u_{j_k}^{(k)} e_{j_k} \right) \\ &= \sum_{j_1=1}^n \sum_{j_2=1}^n \cdots \sum_{j_k=1}^n (D^k f)(a)(e_{j_1}, \dots, e_{j_k}) u_{j_1}^{(1)} u_{j_2}^{(2)} \cdots u_{j_k}^{(k)} \\ &= \sum_{j_1, \dots, j_k=1}^n \frac{\partial^k f}{\partial x_{j_k} \partial x_{j_{k-1}} \cdots \partial x_{j_1}}(a) u_{j_1}^{(1)} u_{j_2}^{(2)} \cdots u_{j_k}^{(k)}. \end{aligned}$$

We prove the proposition by induction. Note that the case $k = 1$ is true because of Theorem 6.21. Next we assume that (6.6.2) holds true for $k = \ell$ if f is $(\ell - 1)$ -times differentiable in a neighborhood of a and f is ℓ -times differentiable at a . Now we show that (6.6.2) also holds true for $k = \ell + 1$ if f is ℓ -times differentiable in a neighborhood of a , and f is $(\ell + 1)$ -times differentiable at a . By the definition of $(\ell + 1)$ -times differentiability at a ,

$$\lim_{x \rightarrow a} \frac{\|(D^\ell f)(x) - (D^\ell f)(a) - (D^{\ell+1} f)(a)(x - a)\|_{\mathcal{B}(\mathbb{R}^n, \mathcal{B}(\mathbb{R}^n, \dots, \mathcal{B}(\mathbb{R}^n, \mathbb{R}) \dots))}}{\|x - a\|_{\mathbb{R}^n}} = 0.$$

Since

$$\begin{aligned} &\left| [(D^\ell f)(x) - (D^\ell f)(a) - (D^{\ell+1} f)(a)(x - a)](e_{j_\ell}) \cdots (e_{j_2})(e_{j_1}) \right| \\ &\leq \left\| [(D^\ell f)(x) - (D^\ell f)(a) - (D^{\ell+1} f)(a)(x - a)](e_{j_\ell}) \cdots (e_{j_2}) \right\|_{\mathcal{B}(\mathbb{R}^n, \mathbb{R})} \|e_{j_1}\|_{\mathbb{R}^n} \\ &\leq \left\| (D^\ell f)(x) - (D^\ell f)(a) - (D^{\ell+1} f)(a)(x - a) \right\|_{\mathcal{B}(\mathbb{R}^n, \mathcal{B}(\mathbb{R}^n, \dots, \mathcal{B}(\mathbb{R}^n, \mathbb{R}) \dots))} \|e_{j_1}\|_{\mathbb{R}^n} \cdots \|e_{j_\ell}\|_{\mathbb{R}^n} \\ &= \left\| (D^\ell f)(x) - (D^\ell f)(a) - (D^{\ell+1} f)(a)(x - a) \right\|_{\mathcal{B}(\mathbb{R}^n, \mathcal{B}(\mathbb{R}^n, \dots, \mathcal{B}(\mathbb{R}^n, \mathbb{R}) \dots))}, \end{aligned}$$

using (6.6.2) (for the case $k = \ell$) we conclude that

$$\begin{aligned} & \lim_{x \rightarrow a} \frac{\left| \frac{\partial^\ell f}{\partial x_{j_\ell} \partial x_{j_{\ell-1}} \cdots \partial x_{j_1}}(x) - \frac{\partial^\ell f}{\partial x_{j_\ell} \partial x_{j_{\ell-1}} \cdots \partial x_{j_1}}(a) - (D^{\ell+1}f)(a)(e_{j_1}, \dots, e_{j_\ell}, x - a) \right|}{\|x - a\|_{\mathbb{R}^n}} \\ &= \lim_{x \rightarrow a} \frac{\left| (D^\ell f)(x)(e_{j_1}, \dots, e_{j_\ell}) - (D^\ell f)(a)(e_{j_1}, \dots, e_{j_\ell}) - (D^{\ell+1}f)(a)(x - a)(e_{j_1}, \dots, e_{j_\ell}) \right|}{\|x - a\|_{\mathbb{R}^n}} \\ &\leq \lim_{x \rightarrow a} \frac{\|(D^\ell f)(x) - (D^\ell f)(a) - (D^{\ell+1}f)(a)(x - a)\|_{\mathcal{B}(\mathbb{R}^n, \mathcal{B}(\mathbb{R}^n, \dots, \mathcal{B}(\mathbb{R}^n, \mathbb{R}) \dots))}}{\|x - a\|_{\mathbb{R}^n}} = 0. \end{aligned}$$

In particular, if $x = a + te_{j_{\ell+1}}$ for some $j_{\ell+1} = 1, \dots, n$, by the definition of partial derivatives we conclude that

$$\begin{aligned} (D^{\ell+1}f)(a)(e_{j_1}, \dots, e_{j_\ell}, e_{j_{\ell+1}}) &= \lim_{t \rightarrow 0} \frac{\frac{\partial^\ell f}{\partial x_{j_\ell} \partial x_{j_{\ell-1}} \cdots \partial x_{j_1}}(a + te_{j_{\ell+1}}) - \frac{\partial^\ell f}{\partial x_{j_\ell} \partial x_{j_{\ell-1}} \cdots \partial x_{j_1}}(a)}{t} \\ &= \frac{\partial^{\ell+1}f}{\partial x_{j_{\ell+1}} \partial x_{j_\ell} \partial x_{j_{\ell-1}} \cdots \partial x_{j_1}}(a) \end{aligned}$$

which is (6.6.2) for the case $k = \ell + 1$. □

Example 6.75. Let $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ be given by $f(x_1, x_2) = x_1^2 \cos x_2$, and $u^{(1)} = (2, 0)$, $u^{(2)} = (1, 1)$, $u^{(3)} = (0, -1)$. Suppose that f is three-times differentiable at $a = (0, 0)$ (in fact it is, but we have not talked about this yet). Then

$$\begin{aligned} (D^3f)(a)(u^{(1)}, u^{(2)}, u^{(3)}) &= \sum_{i,j,k=1}^2 \frac{\partial^3 f}{\partial x_k \partial x_j \partial x_i}(a) u_i^{(1)} u_j^{(2)} u_k^{(3)} = \sum_{j=1}^2 \frac{\partial^3 f}{\partial x_2 \partial x_j \partial x_1}(a) \cdot 2 \cdot u_j^{(2)} \cdot (-1) \\ &= \frac{\partial^3 f}{\partial x_2 \partial x_1^2}(0, 0) \cdot 2 \cdot 1 \cdot (-1) + \frac{\partial^3 f}{\partial x_2^2 \partial x_1}(0, 0) \cdot 2 \cdot 1 \cdot (-1) = 0. \end{aligned}$$

Corollary 6.76. Let $\mathcal{U} \subseteq \mathbb{R}^n$ be open, and $f : \mathcal{U} \rightarrow \mathbb{R}$ be $(k + 1)$ -times differentiable at a . Then for $u^{(1)}, \dots, u^{(k)}, u^{(k+1)} \in \mathbb{R}^n$,

$$(D^{k+1}f)(a)(u^{(1)}, \dots, u^{(k)}, u^{(k+1)}) = \sum_{j=1}^n u_j^{(k+1)} \frac{\partial}{\partial x_j} \Big|_{x=a} (D^k f)(x)(u^{(1)}, \dots, u^{(k)}).$$

In other words, (using the terminology in Remark 6.58) $(D^{k+1}f)(a)(u^{(1)}, \dots, u^{(k)}, u^{(k+1)})$ is the “directional derivative” of the function $(D^k f)(\cdot)(u^{(1)}, \dots, u^{(k)})$ at a in the “direction” $u^{(k+1)}$.

Proof. By Proposition 6.74,

$$\begin{aligned}
 (D^{k+1}f)(a)(u^{(1)}, \dots, u^{(k)}, u^{(k+1)}) &= \sum_{j_1, \dots, j_k, j_{k+1}=1}^n \frac{\partial^{k+1}f}{\partial x_{j_{k+1}} \partial x_{j_k} \dots \partial x_{j_1}}(a) u_{j_1}^{(1)} \dots u_{j_k}^{(k)} u_{j_{k+1}}^{(k+1)} \\
 &= \sum_{j_{k+1}=1}^n u_{j_{k+1}}^{(k+1)} \sum_{j_1, \dots, j_k=1}^n \frac{\partial^{k+1}f}{\partial x_{j_{k+1}} \partial x_{j_k} \dots \partial x_{j_1}}(a) u_{j_1}^{(1)} \dots u_{j_k}^{(k)} \\
 &= \sum_{j_{k+1}=1}^n u_{j_{k+1}}^{(k+1)} \frac{\partial}{\partial x_{j_{k+1}}} \Big|_{x=a} \sum_{j_1, \dots, j_k=1}^n \frac{\partial^k f}{\partial x_{j_k} \dots \partial x_{j_1}}(x) u_{j_1}^{(1)} \dots u_{j_k}^{(k)} \\
 &= \sum_{j_{k+1}=1}^n u_{j_{k+1}}^{(k+1)} \frac{\partial}{\partial x_{j_{k+1}}} \Big|_{x=a} (D^k f)(x)(u^{(1)}, \dots, u^{(k)}). \quad \square
 \end{aligned}$$

Example 6.77. Let $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ be twice differentiable at $a = (a_1, a_2) \in \mathbb{R}^2$. Then the proposition above suggests that for $u = (u_1, u_2), v = (v_1, v_2) \in \mathbb{R}^2$,

$$\begin{aligned}
 (D^2 f)(a)(v)(u) &= (D^2 f)(a)(u, v) = \sum_{i,j=1}^2 \frac{\partial^2 f}{\partial x_j \partial x_i}(a) u_i v_j \\
 &= \frac{\partial^2 f}{\partial x_1^2}(a) u_1 v_1 + \frac{\partial^2 f}{\partial x_2 \partial x_1}(a) u_1 v_2 + \frac{\partial^2 f}{\partial x_1 \partial x_2}(a) u_2 v_1 + \frac{\partial^2 f}{\partial x_2^2}(a) u_2 v_2 \\
 &= [u_1 \quad u_2] \begin{bmatrix} \frac{\partial^2 f}{\partial x_1^2}(a) & \frac{\partial^2 f}{\partial x_2 \partial x_1}(a) \\ \frac{\partial^2 f}{\partial x_1 \partial x_2}(a) & \frac{\partial^2 f}{\partial x_2^2}(a) \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix}.
 \end{aligned}$$

In general, if $f : \mathbb{R}^n \rightarrow \mathbb{R}$ be twice differentiable at $a = (a_1, \dots, a_n) \in \mathbb{R}^n$. Then for $u = (u_1, \dots, u_n), v = (v_1, \dots, v_n) \in \mathbb{R}^n$

$$(D^2 f)(a)(v)(u) = [u_1 \quad \dots \quad u_n] \begin{bmatrix} \frac{\partial^2 f}{\partial x_1^2}(a) & \dots & \frac{\partial^2 f}{\partial x_n \partial x_1}(a) \\ \vdots & \ddots & \vdots \\ \frac{\partial^2 f}{\partial x_1 \partial x_n}(a) & \dots & \frac{\partial^2 f}{\partial x_n^2}(a) \end{bmatrix} \begin{bmatrix} v_1 \\ \vdots \\ v_n \end{bmatrix}.$$

The bilinear form $B : \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R}$ given by

$$B(u, v) = (D^2 f)(a)(v)(u) \quad \forall u, v \in \mathbb{R}^n$$

is called the **Hessian** of f , and is represented (in the matrix form) as an $n \times n$ matrix by

$$\begin{bmatrix} \frac{\partial^2 f}{\partial x_1^2}(a) & \cdots & \frac{\partial^2 f}{\partial x_n \partial x_1}(a) \\ \vdots & \ddots & \vdots \\ \frac{\partial^2 f}{\partial x_1 \partial x_n}(a) & \cdots & \frac{\partial^2 f}{\partial x_n^2}(a) \end{bmatrix}.$$

If the second partial derivatives $\frac{\partial^2 f}{\partial x_j \partial x_i}(a)$ of f at a exists for all $i, j = 1, \dots, n$ (here the twice differentiability of f at a is ignored), the matrix (on the right-hand side of equality) above is also called the **Hessian matrix** of f at a .

Even though there is no reason to believe that $(D^2 f)(a)(u, v) = (D^2 f)(a)(v, u)$ (since the left-hand side means first differentiating f in u -direction and then differentiating Df in v -direction, while the right-hand side means first differentiating f in v -direction then differentiating Df in u -direction), it is still reasonable to ask whether $(D^2 f)(a)$ is symmetric or not; that is, could it be true that $(D^2 f)(a)(u, v) = (D^2 f)(a)(v, u)$ for all $u, v \in \mathbb{R}^n$? When f is twice differentiable at a , this is equivalent of asking (by plugging in $u = e_i$ and $v = e_j$) that whether or not

$$\frac{\partial^2 f}{\partial x_j \partial x_i}(a) = \frac{\partial^2 f}{\partial x_i \partial x_j}(a). \quad (6.6.3)$$

The following example provides a function $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ such that (6.6.3) does not hold at $a = (0, 0)$. We remark that the function in the following example is not twice differentiable at a even though the Hessian matrix of f at a can still be computed.

Example 6.78. Let $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ be defined by

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

Then

$$f_x(x, y) = \begin{cases} \frac{x^4 y + 4x^2 y^3 - y^5}{(x^2 + y^2)^2} & \text{if } (x, y) \neq (0, 0), \\ 0 & \text{if } (x, y) = (0, 0), \end{cases}$$

and

$$f_y(x, y) = \begin{cases} \frac{x^5 - 4x^3 y^2 - xy^4}{(x^2 + y^2)^2} & \text{if } (x, y) \neq (0, 0), \\ 0 & \text{if } (x, y) = (0, 0), \end{cases}$$

It is clear that f_x and f_y are continuous on \mathbb{R}^2 ; thus f is differentiable on \mathbb{R}^2 . However,

$$f_{xy}(0, 0) = \lim_{k \rightarrow 0} \frac{f_x(0, k) - f_x(0, 0)}{k} = -1,$$

while

$$f_{yx}(0, 0) = \lim_{h \rightarrow 0} \frac{f_y(h, 0) - f_y(0, 0)}{h} = 1;$$

thus the Hessian matrix of f at the origin is not symmetric.

Definition 6.79. A function is said to be **of class** \mathcal{C}^r if the first r derivatives exist and are continuous. A function is said to be **smooth** or **of class** \mathcal{C}^∞ if it is of class \mathcal{C}^r for all positive integer r .

Now we would like to answer the question of what kind of functions are k -times differentiable. Suppose that $\mathcal{U} \subseteq \mathbb{R}^n$ is open and $f : \mathcal{U} \rightarrow \mathbb{R}$. Note that by the definition of differentiability, f is k -times differentiable in \mathcal{U} if and only if $D^{k-1}f$ is differentiable in \mathcal{U} . This would further imply that f is k -times differentiable in \mathcal{U} if and only if $D^{k-2}f$ is twice differentiable in \mathcal{U} . Therefore, Proposition 6.27 and Corollary 6.32 imply that

$$\begin{aligned} & f \text{ is } k\text{-times (continuously) differentiable in } \mathcal{U} \\ & \Leftrightarrow Df \text{ is } (k-1)\text{-times (continuously) differentiable in } \mathcal{U} \\ & \Leftrightarrow \left[\frac{\partial f}{\partial x_1}, \frac{\partial f}{\partial x_2}, \dots, \frac{\partial f}{\partial x_n} \right] \text{ is } (k-1)\text{-times (continuously) differentiable in } \mathcal{U} \\ & \Leftrightarrow \frac{\partial f}{\partial x_{j_1}} \text{ is } (k-1)\text{-times (continuously) differentiable in } \mathcal{U} \text{ for all } 1 \leq j_1 \leq n \\ & \Leftrightarrow D \frac{\partial f}{\partial x_{j_1}} \text{ is } (k-2)\text{-times (continuously) differentiable in } \mathcal{U} \text{ for all } 1 \leq j_1 \leq n \\ & \Leftrightarrow \left[\frac{\partial^2 f}{\partial x_1 \partial x_{j_1}}, \dots, \frac{\partial^2 f}{\partial x_n \partial x_{j_1}} \right] \text{ is } (k-2)\text{-times (continuously) differentiable in } \mathcal{U} \\ & \quad \text{for all } 1 \leq j_1 \leq n \\ & \Leftrightarrow \frac{\partial^2 f}{\partial x_{j_2} \partial x_{j_1}} \text{ is } (k-2)\text{-times (continuously) differentiable in } \mathcal{U} \text{ for all } 1 \leq j_1, j_2 \leq n. \end{aligned}$$

Applying similar argument several times, we obtain the following theorem which is an analogy of Corollary 6.32.

Theorem 6.80. Let $\mathcal{U} \subseteq \mathbb{R}^n$ and $f : \mathcal{U} \rightarrow \mathbb{R}$. Suppose that the partial derivative $\frac{\partial^k f}{\partial x_{j_k} \partial x_{j_{k-1}} \cdots \partial x_{j_1}}$ exists in a neighborhood of $a \in \mathcal{U}$ and is continuous at a for all $j_1, \dots, j_k =$

$1, \dots, n$. Then f is k -times differentiable at a . Moreover, if $\frac{\partial^k f}{\partial x_{j_k} \partial x_{j_{k-1}} \cdots \partial x_{j_1}}$ is continuous on \mathcal{U} , then f is of class \mathcal{C}^k .

Theorem 6.81. Let $\mathcal{U} \subseteq \mathbb{R}^n$ be open, and $f : \mathcal{U} \rightarrow \mathbb{R}$. Suppose that the mixed partial derivatives $\frac{\partial f}{\partial x_i}$, $\frac{\partial f}{\partial x_j}$, $\frac{\partial^2 f}{\partial x_j \partial x_i}$, $\frac{\partial^2 f}{\partial x_i \partial x_j}$ exist in a neighborhood of a , and are continuous at a . Then

$$\frac{\partial^2 f}{\partial x_j \partial x_i}(a) = \frac{\partial^2 f}{\partial x_i \partial x_j}(a). \quad (6.6.4)$$

Proof. Let $S(a, h, k) = f(a + he_i + ke_j) - f(a + he_i) - f(a + ke_j) + f(a)$, and define $\varphi(x) = f(x + he_i) - f(x)$ as well as $\psi(x) = f(x + ke_j) - f(x)$ for x in a neighborhood of a . Then $S(a, h, k) = \varphi(a + ke_j) - \varphi(a) = \psi(a + he_i) - \psi(a)$; thus the mean value theorem implies that there exists c on the line segment joining a and $a + ke_j$ and d on the line segment joining a and $a + he_i$ such that

$$\begin{aligned} S(a, h, k) &= \varphi(a + ke_j) - \varphi(a) = k \frac{\partial \varphi}{\partial x_j}(c) = k \left(\frac{\partial f}{\partial x_j}(c + he_i) - \frac{\partial f}{\partial x_j}(c) \right), \\ S(a, h, k) &= \psi(a + he_i) - \psi(a) = h \frac{\partial \psi}{\partial x_i}(d) = h \left(\frac{\partial f}{\partial x_i}(d + ke_j) - \frac{\partial f}{\partial x_i}(d) \right). \end{aligned}$$

As a consequence, if $h \neq 0 \neq k$,

$$\frac{1}{k} \left(\frac{\partial f}{\partial x_i}(d + ke_j) - \frac{\partial f}{\partial x_i}(d) \right) = \frac{S(a, h, k)}{hk} = \frac{1}{h} \left(\frac{\partial f}{\partial x_j}(c + he_i) - \frac{\partial f}{\partial x_j}(c) \right)$$

By the mean value theorem again, there exists c_1 and d_1 on the line segment joining c , $c + he_i$ and d , $d + ke_j$, respectively, such that

$$\frac{\partial^2 f}{\partial x_j \partial x_i}(d_1) = \frac{\partial^2 f}{\partial x_i \partial x_j}(c_1).$$

The theorem is then concluded by the continuity of $\frac{\partial^2 f}{\partial x_i \partial x_j}$ and $\frac{\partial^2 f}{\partial x_j \partial x_i}$ at a , and $c_1 \rightarrow a$ and $d_1 \rightarrow a$ as $(h, k) \rightarrow (0, 0)$. \square

Corollary 6.82. Let $\mathcal{U} \subseteq \mathbb{R}^n$ be open, and f is of class \mathcal{C}^2 . Then

$$(D^2 f)(a)(u, v) = (D^2 f)(a)(v, u) \quad \forall a \in \mathcal{U} \text{ and } u, v \in \mathbb{R}^n.$$

Remark 6.83. In view of Remark 6.70, (6.6.4) is the same as the following identity

$$\begin{aligned} & \lim_{h \rightarrow 0} \lim_{k \rightarrow 0} \frac{f(a + he_i + ke_j) - f(a + he_i) - f(a + ke_j) + f(a)}{hk} \\ &= \lim_{k \rightarrow 0} \lim_{h \rightarrow 0} \frac{f(a + he_i + ke_j) - f(a + he_i) - f(a + ke_j) + f(a)}{hk} \end{aligned}$$

which implies that the order of the two limits $\lim_{h \rightarrow 0}$ and $\lim_{k \rightarrow 0}$ can be interchanged without changing the value of the limit (under certain conditions).

Example 6.84. Let $f(x, y) = yx^2 \cos y^2$. Then

$$\begin{aligned} f_{xy}(x, y) &= (2xy \cos y^2)_y = 2x \cos y^2 - 2xy(2y) \sin y^2 = 2x \cos y^2 - 4xy^2 \sin y^2, \\ f_{yx}(x, y) &= (x^2 \cos y^2 - yx^2(2y) \sin y^2)_x = (x^2 \cos y^2 - 2x^2 y^2 \sin y^2)_x \\ &= 2x \cos y^2 - 4xy^2 \sin y^2 = f_{xy}(x, y). \end{aligned}$$

The following two theorems concern the \mathcal{C}^k -regularity of inverse functions and implicit functions.

Theorem 6.85. Let $\mathcal{D} \subseteq \mathbb{R}^n$ be open, $f : \mathcal{D} \rightarrow \mathbb{R}^n$ be injective and be of class \mathcal{C}^k . If f^{-1} , the inverse function of f , exists and is differentiable in $f(\mathcal{D})$, then f^{-1} is of class \mathcal{C}^k .

Proof. Let $y_0 \in f(\mathcal{D})$. Then $y_0 = f(x_0)$ for some $x_0 \in \mathcal{D}$. Since f is differentiable at x_0 and f^{-1} is differentiable at y_0 , by the chain rule we must have

$$I_n = [D(f \circ f^{-1})](y_0) = [Df](x_0)[Df^{-1}](y_0),$$

where I_n is the $n \times n$ identity matrix. Therefore, $[Df](x_0)$ is invertible, and the inverse function theorem implies that f^{-1} is of class \mathcal{C}^1 (in a neighborhood of y_0).

We note that the map $g : \text{GL}(n) \rightarrow \text{GL}(n)$ given by $g(L) = L^{-1}$ is infinitely many times differentiable; thus using the identity (from the inverse function theorem)

$$(Df^{-1})(y) = ((Df)(x))^{-1} = (g \circ (Df) \circ f^{-1})(y),$$

by the chain rule we find that if $f \in \mathcal{C}^k$, then $Df^{-1} \in \mathcal{C}^{k-1}$ which is the same as saying that $f^{-1} \in \mathcal{C}^k$. □

Theorem 6.86. Let $\mathcal{D} \subseteq \mathbb{R}^n \times \mathbb{R}^m$ be open, and $F : \mathcal{D} \rightarrow \mathbb{R}^m$ be a function of class \mathcal{C}^k . Suppose that for some open set $\mathcal{U} \subseteq \mathbb{R}^n$ and some differentiable function $f : \mathcal{U} \rightarrow \mathbb{R}^m$, $\mathcal{U} \times f(\mathcal{U}) \subseteq \mathcal{D}$ and $F(x, f(x)) = 0$ for all $x \in \mathcal{U}$. Then f is of class \mathcal{C}^k .

Proof. **(Not yet finished!!!)** □

6.7 Taylor's Theorem

Recall the Taylor theorem for functions of one variable that if $f : (a, b) \rightarrow \mathbb{R}$ be of class \mathcal{C}^{k+1} for some $k \in \mathbb{N}$ and $c \in (a, b)$, then for all $x \in (a, b)$, there exists d in between c and x such that

$$f(x) = \sum_{j=0}^k \frac{f^{(j)}(c)}{j!} (x - c)^j + \frac{f^{(k+1)}(d)}{(k+1)!} (x - c)^{k+1},$$

where $f^{(j)}(c)$ denotes the j -th derivative of f at c . In this section, we extend this result to functions of several variables.

Theorem 6.87 (Taylor). *Let $\mathcal{U} \subseteq \mathbb{R}^n$ be open, and $f : \mathcal{U} \rightarrow \mathbb{R}$ be $(k+1)$ -times differentiable. Suppose that $x, a \in \mathcal{U}$ and the line segment joining x and a lies in \mathcal{U} . Then there exists a point c on the line segment joining x and a such that*

$$f(x) - f(a) = \sum_{j=1}^k \frac{1}{j!} (D^j f)(a) \underbrace{(x - a, \dots, x - a)}_{j \text{ copies of } x - a} + \frac{1}{(k+1)!} (D^{k+1} f)(c) \underbrace{(x - a, \dots, x - a)}_{(k+1) \text{ copies of } x - a}. \tag{6.7.1}$$

Proof. Let $g(t) = f((1-t)a + tx)$. Since $\overline{xa} \subseteq \mathcal{U}$ and \mathcal{U} is open, there exists $\delta > 0$ such that $(1-t)a + tx \in \mathcal{U}$ for all $t \in (-\delta, 1 + \delta)$. By the chain rule, for $t \in (-\delta, 1 + \delta)$,

$$g'(t) = (Df)((1-t)a + tx)(x - a) = \sum_{i=1}^n \frac{\partial f}{\partial x_i}((1-t)a + tx)(x_i - a_i);$$

thus for $t \in (-\delta, 1 + \delta)$, Proposition 6.74 shows that

$$g''(t) = \sum_{i,j=1}^n \frac{\partial^2 f}{\partial x_j \partial x_i}((1-t)a + tx)(x_i - a_i)(x_j - a_j) = (D^2 f)((1-t)a + tx)(x - a, x - a).$$

By induction, we conclude that

$$g^{(j)}(t) = (D^j f)((1-t)a + tx) \underbrace{(x - a, \dots, x - a)}_{j \text{ copies of } x - a}.$$

By the fact that f is $(k+1)$ -times differentiable, $g : (-\delta, 1 + \delta) \rightarrow \mathbb{R}$ is $(k+1)$ -times differentiable as well. Theorem 4.68 then implies that for some $t_0 \in (0, 1)$,

$$g(1) - g(0) = \sum_{j=1}^k \frac{g^{(j)}(0)}{j!} + \frac{g^{(k+1)}(t_0)}{(k+1)!}. \tag{6.7.2}$$

Letting $c = (1 - t_0)a + t_0x$, (6.7.2) implies (6.7.1). □

Definition 6.88. Let $\mathcal{U} \subseteq \mathbb{R}^n$ be open, and $f : \mathcal{U} \rightarrow \mathbb{R}$ be k -times differentiable. The k -th degree Taylor polynomial for f centered at a is the polynomial

$$\sum_{j=0}^k \frac{1}{j!} (D^j f)(a) \underbrace{(x - a, \dots, x - a)}_{j \text{ copies } x - a}.$$

Corollary 6.89. Let $\mathcal{U} \subseteq \mathbb{R}^n$ be open, $f : \mathcal{U} \rightarrow \mathbb{R}$ be $(k+1)$ -times differentiable, and define the remainder

$$R_k(a, h) = f(a + h) - \sum_{j=0}^k \frac{1}{j!} (D^j f)(a)(h, \dots, h).$$

Then $\lim_{h \rightarrow 0} \frac{R_k(a, h)}{\|h\|_{\mathbb{R}^n}^k} = 0$, or in notation, $R_k(a, h) = o(\|h\|_{\mathbb{R}^n}^k)$ as $h \rightarrow 0$.

Example 6.90. Let $f(x, y) = e^x \cos y$. Compute the fourth degree Taylor polynomial for f centered at $(0, 0)$.

Solution: We compute the zeroth, the first, the second, the third and the fourth mixed derivatives of f at $(0, 0)$ as follows:

$$\begin{aligned} f(0, 0) &= 1, & f_x(0, 0) &= 1, & f_y(0, 0) &= 0, \\ f_{xx}(0, 0) &= 1, & f_{xy}(0, 0) &= f_{yx}(0, 0) = 0, & f_{yy}(0, 0) &= -1, \\ f_{xxx}(0, 0) &= 1, & f_{xxy}(0, 0) &= f_{xyx}(0, 0) = f_{yxx}(0, 0) = 0, \\ f_{yyy}(0, 0) &= 0, & f_{yyx}(0, 0) &= f_{yxy}(0, 0) = f_{xyy}(0, 0) = -1, \end{aligned}$$

and

$$\begin{aligned} f_{xxxx}(0, 0) &= 1, & f_{yyyy}(0, 0) &= 1, \\ f_{xxxxy}(0, 0) &= f_{xxyx}(0, 0) = f_{xyxx}(0, 0) = f_{yxxx}(0, 0) = 0, \\ f_{xyyyy}(0, 0) &= f_{yxyy}(0, 0) = f_{yyxy}(0, 0) = f_{yyyx}(0, 0) = 0, \\ f_{xxyyy}(0, 0) &= f_{xyxy}(0, 0) = f_{xyyx}(0, 0) = f_{yxyx}(0, 0) \\ &= f_{yxxy}(0, 0) = f_{yyxx}(0, 0) = -1. \end{aligned}$$

Then the fourth degree Taylor polynomial for f centered at $(0, 0)$ is

$$\begin{aligned} & f(0, 0) + f_x(0, 0)x + f_y(0, 0)y + \frac{1}{2} \left[f_{xx}(0, 0)x^2 + 2f_{xy}(0, 0)xy + f_{yy}(0, 0)y^2 \right] \\ & + \frac{1}{6} \left[f_{xxx}(0, 0)x^3 + 3f_{xxy}(0, 0)x^2y + 3f_{xyy}(0, 0)xy^2 + f_{yyy}(0, 0)y^3 \right] \\ & + \frac{1}{24} \left[f_{xxxx}(0, 0)x^4 + 4f_{xxxxy}(0, 0)x^3 + 6f_{xxyy}(0, 0)x^2y^2 \right. \\ & \quad \left. + 4f_{xyyy}(0, 0)xy^3 + f_{yyyy}(0, 0)y^4 \right] \\ & = 1 + x + \frac{1}{2}(x^2 - y^2) + \frac{1}{6}(x^3 - 3xy^2) + \frac{1}{24}(x^4 - 6x^2y^2 + y^4). \end{aligned}$$

Observing that using the Taylor expansions

$$e^x = 1 + x + \frac{1}{2}x^2 + \frac{1}{6}x^3 + \frac{1}{24}x^4 + \cdots \quad \text{and} \quad \cos y = 1 - \frac{1}{2}y^2 + \frac{1}{24}y^4 + \cdots,$$

we can “formally” compute $e^x \cos y$ by multiplying the two “polynomials” above and obtain that

$$e^x \cos y \text{ “=” } 1 + x + \frac{1}{2}(x^2 - y^2) + \left(\frac{1}{6}x^3 - \frac{1}{2}xy^2\right) + \left(\frac{1}{24}x^4 - \frac{1}{4}x^2y^2 + \frac{1}{24}y^2\right) + \text{h.o.t.};$$

where h.o.t. stands for the higher order terms which are terms with fifth or higher degree.

Definition 6.91. Let $\mathcal{U} \subseteq \mathbb{R}^n$ be open. A function $f : \mathcal{U} \rightarrow \mathbb{R}$ is said to be **real analytic** at $a \in \mathcal{U}$ if $f(x) = \sum_{k=0}^{\infty} \frac{1}{k!} (D^k f)(a)(x - a, \dots, x - a)$ in a neighborhood of a .

Example 6.92. Let $f : \mathbb{R} \rightarrow \mathbb{R}$ be defined by

$$f(x) = \begin{cases} \exp\left(-\frac{1}{|x|^2}\right) & \text{if } x > 0, \\ 0 & \text{if } x < 0. \end{cases}$$

Then f is of class \mathcal{C}^∞ , and $f^{(k)}(0) = 0$ for all $k \in \mathbb{N}$. Therefore, f is not real analytic at 0.

6.8 Maxima and Minima

Definition 6.93. Let $\mathcal{U} \subseteq \mathbb{R}^n$ be open, and $f : \mathcal{U} \rightarrow \mathbb{R}$.

1. If there is a neighborhood of $x_0 \in \mathcal{U}$ such that $f(x_0)$ is a maximum in this neighborhood, then x_0 is called a **local maximum point** of f .

2. If there is a neighborhood of $x_0 \in \mathcal{U}$ such that $f(x_0)$ is a minimum in this neighborhood, then x_0 is called a **local minimum point** of f .
3. A point is called an **extreme point** of f if it is either a local maximum point or a local minimum point of f .
4. A point x_0 is a **critical point** of f if f is differentiable at x_0 and $(Df)(x_0) = 0$; that is, $(Df)(x_0) \in \mathcal{B}(\mathbb{R}^n, \mathbb{R})$ is the trivial map (which sends every vector in \mathbb{R}^n to zero vector).
5. A point x_0 is a **saddle point** of f if x_0 is a critical point of f but not an extreme point of f .

Theorem 6.94. Let $\mathcal{U} \subseteq \mathbb{R}^n$ be open, $f : \mathcal{U} \rightarrow \mathbb{R}$ be differentiable, and $x_0 \in \mathcal{U}$ is an extreme point of f . Then x_0 is a critical point of f .

Proof. Suppose the contrary that the linear map $(Df)(x_0) : \mathbb{R}^n \rightarrow \mathbb{R}$ is not the zero map; that is, there exists $u \in \mathbb{R}^n$, $u \neq 0$, such that $(Df)(x_0)(u) = c \neq 0$ for some constant $c \in \mathbb{R}$. W.L.O.G, we can assume that $\|u\|_{\mathbb{R}^n} = 1$ and $c > 0$ (for otherwise change u to $-u$). By the differentiability of f ,

$$\exists \delta > 0 \ni |f(x_0 + h) - f(x_0) - (Df)(x_0)(h)| \leq \frac{c}{2} \|h\|_{\mathbb{R}^n} \quad \text{whenever } \|h\|_{\mathbb{R}^n} < \delta.$$

Then for any $0 < \lambda < \delta$,

$$|f(x_0 \pm \lambda u) - f(x_0) \mp \lambda(Df)(x_0)(u)| \leq \frac{\lambda c}{2}.$$

Therefore, $-\frac{\lambda c}{2} \leq f(x_0 \pm \lambda u) - f(x_0) \mp \lambda c \leq \frac{\lambda c}{2}$ which further implies that

$$f(x_0) \leq f(x_0 + \lambda u) - \frac{\lambda c}{2} < f(x_0 + \lambda u) \quad \text{and} \quad f(x_0) \geq f(x_0 - \lambda u) + \frac{\lambda c}{2} > f(x_0 - \lambda u)$$

for all $\lambda > 0$ small enough. As a consequence, x_0 cannot be a local extreme point of f , a contradiction. \square

Definition 6.95. If $f : \mathcal{U} \rightarrow \mathbb{R}$ is of class \mathcal{C}^2 , the **Hessian of f at x_0** is the bilinear function $H_{x_0}(f) : \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R}$ given by

$$H_{x_0}(f)(u, v) = (D^2f)(x_0)(u, v) \quad \forall u, v \in \mathbb{R}^n.$$

The matrix representation of $H_{x_0}(f)(\cdot, \cdot)$ is given by

$$[H_{x_0}(f)] = \begin{bmatrix} \frac{\partial^2 f}{\partial x_1^2}(x_0) & \cdots & \frac{\partial^2 f}{\partial x_n \partial x_1}(x_0) \\ \vdots & \ddots & \vdots \\ \frac{\partial^2 f}{\partial x_1 \partial x_n}(x_0) & \cdots & \frac{\partial^2 f}{\partial x_n^2}(x_0) \end{bmatrix}$$

in the sense that $H_{x_0}(f)(u, v) = [u]^T [H_{x_0}(f)] [v] = [v]^T [H_{x_0}(f)] [u]$.

Definition 6.96. A bilinear form $B : \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R}$ is called **positive definite** if $B(u, u) > 0$ for all $u \neq 0$, and is called **negative definite** if $B(u, u) < 0$ for all $u \neq 0$. It is called **positive semi-definite** if $B(u, u) \geq 0$ for all $u \in \mathbb{R}^n$, and **negative semi-definite** if $B(u, u) \leq 0$ for all $u \in \mathbb{R}^n$.

Theorem 6.97. Let $\mathcal{U} \subseteq \mathbb{R}^n$ be open, and $f : \mathcal{U} \rightarrow \mathbb{R}$ be a function of class \mathcal{C}^2 .

1. If x_0 is a critical point of f such that the Hessian $H_{x_0}(f)$ is **negative definite**, then f has a local **maximum** point at x_0 .
2. If f has a local **maximum** point at x_0 , then $H_{x_0}(f)$ is **negative semi-definite**.

Proof. 1. Suppose that $H_{x_0}(f)$ is negative definite.

Claim: There exists $0 < \lambda < \infty$ such that

$$H_{x_0}(f)(u, u) \leq -\lambda \|u\|_{\mathbb{R}^n}^2 \quad \forall u \in \mathbb{R}^n. \tag{6.8.1}$$

Proof of claim: Since $H_{x_0}(f)(u, u)$, viewed as a function of u , is continuous, by Theorem 4.21 $\lambda = -\max_{\|u\|_{\mathbb{R}^n}=1} H_{x_0}(f)(u, u)$ exists and is positive. Then for all $u \in \mathbb{R}^n$ with $u \neq 0$,

$$H_{x_0}(f)\left(\frac{u}{\|u\|_{\mathbb{R}^n}}, \frac{u}{\|u\|_{\mathbb{R}^n}}\right) \leq -\lambda \quad \forall u \in \mathbb{R}^n, u \neq 0.$$

The inequality (6.8.1) follows from that the Hessian $H_{x_0}(f)$ is bilinear.

Since $f \in \mathcal{C}^2$, there exists $\delta > 0$ such that $D(x_0, \delta) \subseteq \mathcal{U}$ and

$$\|(D^2 f)(x) - (D^2 f)(x_0)\|_{\mathcal{B}(\mathbb{R}^n, \mathcal{B}(\mathbb{R}^n, \mathbb{R}))} \leq \frac{\lambda}{2} \quad \forall x \in D(x_0, \delta). \tag{6.8.2}$$

Now since x_0 is a critical point of f , $(Df)(x_0) = 0$. As a consequence, by Taylor's theorem (Theorem 6.87), for any $x \in D(x_0, \delta)$, we can find $c = c(x) \in \overline{xx_0}$ such that

$$\begin{aligned} f(x) &= f(x_0) + (Df)(x_0)(x - x_0) + \frac{1}{2}(D^2f)(c)(x - x_0, x - x_0) \\ &= f(x_0) + \frac{1}{2}(D^2f)(x_0)(x - x_0, x - x_0) + \frac{1}{2}[(D^2f)(c) - (D^2f)(x_0)](x - x_0, x - x_0) \\ &\leq f(x_0) - \frac{1}{2}\lambda\|x - x_0\|_{\mathbb{R}^n}^2 + \frac{1}{2}\left|[(D^2f)(c) - (D^2f)(x_0)](x - x_0, x - x_0)\right| \\ &\leq f(x_0) - \frac{1}{2}\lambda\|x - x_0\|_{\mathbb{R}^n}^2 + \frac{1}{2}\|(D^2f)(c) - (D^2f)(x_0)\|_{\mathcal{B}(\mathbb{R}^n; \mathcal{B}(\mathbb{R}^n, \mathbb{R}))}\|x - x_0\|_{\mathbb{R}^n}^2. \end{aligned}$$

Note that $c = c(x) \in D(x_0, \delta)$ if $x \in D(x_0, \delta)$; thus (6.8.2) implies that if $x \in D(x_0, \delta)$,

$$f(x) \leq f(x_0) - \frac{\lambda}{2}\|x - x_0\|_{\mathbb{R}^n}^2 + \frac{1}{2}\frac{\lambda}{2}\|x - x_0\|_{\mathbb{R}^n}^2 \leq f(x_0) - \frac{\lambda}{4}\|x - x_0\|_{\mathbb{R}^n}^2.$$

As a consequence, for all $x \in D(x_0, \delta)$, $f(x) \leq f(x_0)$ which validates that x_0 is a local maximum point of f .

2. Suppose the contrary that f has a local maximum point at x_0 but for some $u \in \mathbb{R}^n$,

$$H_{x_0}(f)(u, u) > 0.$$

W.L.O.G, we can assume that $\|u\|_{\mathbb{R}^n} = 1$. By Theorem 6.94, $(Df)(x_0) = 0$; thus Taylor's Theorem implies that

$$f(x) = f(x_0) + \frac{1}{2}(D^2f)(c)(x - x_0, x - x_0) = f(x_0) + \frac{1}{2}(x - x_0)^T [H_c(f)](x - x_0).$$

Since x_0 is a local maximum point of f , there exists $\delta > 0$ such that $f(x) \leq f(x_0)$ for all $x \in D(x_0, \delta)$. As a consequence, for some $c = c(x) \in \overline{xx_0}$,

$$(x - x_0)^T [H_c(f)](x - x_0) = 2[f(x) - f(x_0)] \leq 0 \quad \forall x \in D(x_0, \delta).$$

Let $0 < t < \delta$ and $x = x_0 + tu$. Then $x \in D(x_0, \delta)$; thus

$$H_c(f)(u, u) \leq 0 \quad \forall t \in (0, \delta).$$

We note that c depends on t , and $c \rightarrow x_0$ as $t \rightarrow 0$. Therefore, by the continuity of $H_\bullet(f)$, passing $t \rightarrow 0$ in the inequality above we find that

$$H_{x_0}(f)(u, u) = \lim_{t \rightarrow 0} H_c(f)(u, u) \leq 0$$

which is a contradiction. □

Remark 6.98. Inequality (6.8.1) can also be obtained by studying the largest eigenvalue of $H_{x_0}(f)$. Note that since $f \in \mathcal{C}^2$, $H_{x_0}(f)$ is symmetric by Theorem 6.81. As a consequence, there exists an orthonormal matrix $\mathbb{O} \in \text{GL}(n)$ whose columns are (real) eigenvectors of $H_{x_0}(f)$

$$[H_{x_0}(f)] = \mathbb{O}\Lambda\mathbb{O}^T,$$

where Λ is a diagonal matrix whose diagonal entries are eigenvalues of $H_{x_0}(f)$. Note that by the orthonormality of \mathbb{O} , every vector $u \in \mathbb{R}^n$ satisfies $\|\mathbb{O}^T u\|_{\mathbb{R}^n} = \|u\|_{\mathbb{R}^n}$. Therefore,

$$H_{x_0}(f)(u, u) = u^T \mathbb{O} \Lambda \mathbb{O}^T u = (\mathbb{O}^T u)^T \Lambda (\mathbb{O}^T u) \leq \lambda \|\mathbb{O}^T u\|_{\mathbb{R}^n}^2 = \lambda \|u\|_{\mathbb{R}^n}^2,$$

where λ is the largest eigenvalue of Λ .

Remark 6.99 (Sylvester's criterion). To justify if a matrix $[H_{x_0}(f)]$ is positive/negative definite, let

$$\Delta_k = \begin{bmatrix} \frac{\partial^2 f}{\partial x_1^2} & \cdots & \frac{\partial^2 f}{\partial x_k \partial x_1} \\ \vdots & \ddots & \vdots \\ \frac{\partial^2 f}{\partial x_1 \partial x_k} & \cdots & \frac{\partial^2 f}{\partial x_k^2} \end{bmatrix} (x_0).$$

Then $H_{x_0}(f)$ is $\begin{matrix} \text{positive} \\ \text{negative} \end{matrix}$ definite if and only if $\begin{matrix} \det(\Delta_k) > 0 \\ (-1)^k \det(\Delta_k) > 0 \end{matrix}$ for all $k = 1, \dots, n$.

6.9 Exercises

Problem 6.1. Let $\{T_k\}_{k=1}^\infty \subseteq \mathcal{B}(\mathbb{R}^n, \mathbb{R}^m)$ be a sequence of bounded linear maps from $\mathbb{R}^n \rightarrow \mathbb{R}^m$. Prove that the following three statements are equivalent:

1. $\{T_k\}_{k=1}^\infty$ converges pointwise (to a function T);
2. $\lim_{k, \ell \rightarrow \infty} \|T_k - T_\ell\|_{\mathcal{B}(\mathbb{R}^n, \mathbb{R}^m)} = 0$;
3. $\{T_k\}_{k=1}^\infty$ converges uniformly (to T) on every compact subsets of \mathbb{R}^n .

Problem 6.2. Let $\mathcal{P}((0, 1)) \subseteq \mathcal{C}_b((0, 1); \mathbb{R})$ be the collection of all polynomials defined on $(0, 1)$.

1. Show that the operator $\frac{d}{dx} : \mathcal{P}((0, 1)) \rightarrow \mathcal{C}_b((0, 1))$ is linear.

2. Show that $\frac{d}{dx} : (\mathcal{P}((0,1)), \|\cdot\|_\infty) \rightarrow (\mathcal{C}_b((0,1)), \|\cdot\|_\infty)$ is unbounded; that is, show that

$$\sup_{\|p\|_\infty=1} \|p'\|_\infty = \infty.$$

§6.2 Definition of Derivatives and the Jacobian Matrices

Problem 6.3. Show that if $f : \mathbb{C} \rightarrow \mathbb{R}$ is differentiable at z_0 , then $(Df)(z_0) = 0$.

Hint: Show that $\mathcal{B}(\mathbb{C}, \mathbb{R}) = \{0\}$.

Problem 6.4. Let $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ be given by

$$f(x, y) = \begin{cases} 0 & \text{if } xy = 0, \\ 1 & \text{if } xy \neq 0. \end{cases}$$

Compute $\frac{\partial f}{\partial x}(x, y)$ and $\frac{\partial f}{\partial y}(x, y)$.

Problem 6.5. Investigate the differentiability of

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

Problem 6.6. Investigate the differentiability of

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

Problem 6.7. Define $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ by

$$f(x, y) = \begin{cases} (x^2 + y^2) \sin \frac{1}{\sqrt{x^2 + y^2}} & \text{if } (x, y) \neq (0, 0), \\ 0 & \text{if } (x, y) = (0, 0). \end{cases}$$

Discuss the differentiability of f . Find $(\nabla f)(x, y)$ at points of differentiability.

Problem 6.8. Let $r > 0$ and $\alpha > 1$. Suppose that $f : D(0, r) \rightarrow \mathbb{R}$ satisfies $|f(x)| \leq \|x\|^\alpha$ for all $x \in D(0, r)$. Show that f is differentiable at 0. What happens if $\alpha = 1$?

Problem 6.9. Suppose that $f, g : \mathbb{R} \rightarrow \mathbb{R}^m$ are differentiable at a and there is a $\delta > 0$ such that $g(x) \neq 0$ for all $0 < |x - a| < \delta$. If $f(a) = g(a) = 0$ and $(Dg)(a) \neq 0$, show that

$$\lim_{x \rightarrow a} \frac{\|f(x)\|}{\|g(x)\|} = \frac{\|(Df)(a)\|}{\|(Dg)(a)\|}.$$

Problem 6.10. Consider the map δ defined in Problem 5.11 in Chapter 5; that is, $\delta : \mathcal{C}([0, 1]; \mathbb{R}) \rightarrow \mathbb{R}$ be defined by $\delta(f) = f(0)$. Show that δ is differentiable. Find $(D\delta)(f)$ (for $f \in \mathcal{C}([0, 1]; \mathbb{R})$).

Problem 6.11. Let $f : \text{GL}(n) \rightarrow \text{GL}(n)$ be given by $f(L) = L^{-1}$. In class we have shown that f is continuous on $\text{GL}(n)$. Show that f is differentiable at each “point” (or more precisely, linear map) of $\text{GL}(n)$.

Hint: In order to show the differentiability of f at $L \in \text{GL}(n)$, we need to figure out what $(Df)(L)$ is. So we need to compute $f(L + h) - f(L)$, where $h \in \mathcal{B}(\mathbb{R}^n, \mathbb{R}^n)$ is a “small” linear map. Compute $(L + h)^{-1} - L^{-1}$ and make a conjecture what $(Df)(L)$ should be.

Problem 6.12. Let $I : \mathcal{C}([0, 1]; \mathbb{R}) \rightarrow \mathbb{R}$ be defined by

$$I(f) = \int_0^1 f(x)^2 dx.$$

Show that I is differentiable at every “point” $f \in \mathcal{C}([0, 1]; \mathbb{R})$.

Hint: Figure out what $(DI)(f)$ is by computing $I(f + h) - I(f)$, where $h \in \mathcal{C}([0, 1]; \mathbb{R})$ is a “small” continuous function.

Remark. A map from a space of functions such as $\mathcal{C}([0, 1]; \mathbb{R})$ to a scalar field such as \mathbb{R} or \mathbb{C} is usually called a **functional**. The derivative of a functional I is usually denoted by δI instead of DI .

§6.3 Conditions for Differentiability

§6.4 Properties of Differentiable Functions

Problem 6.13. Let $\mathcal{U} \subseteq \mathbb{R}^n$ be open, and $f : \mathcal{U} \rightarrow \mathbb{R}$. Suppose that the partial derivatives $\frac{\partial f}{\partial x_1}, \dots, \frac{\partial f}{\partial x_n}$ are bounded on \mathcal{U} ; that is, there exists a real number $M > 0$ such that

$$\left| \frac{\partial f}{\partial x_j}(x) \right| \leq M \quad \forall x \in \mathcal{U} \text{ and } j = 1, \dots, n.$$

Show that f is continuous on \mathcal{U} .

Hint: Mimic the proof of Theorem 6.28.

Problem 6.14. Let $\mathcal{U} \subseteq \mathbb{R}^n$ be open, and $f : \mathcal{U} \rightarrow \mathbb{R}$. Show that f is differentiable at $a \in \mathcal{U}$ if and only if there exists a vector-valued function $\varepsilon : \mathcal{U} \rightarrow \mathbb{R}^n$ such that

$$f(x) - f(a) - \sum_{j=1}^n \frac{\partial f}{\partial x_j}(a)(x_j - a_j) = \varepsilon(x) \cdot (x - a)$$

and $\varepsilon(x) \rightarrow 0$ as $x \rightarrow a$.

Problem 6.15. Verify the chain rule for

$$u(x, y, z) = xe^y, \quad v(x, y, z) = yz \sin x$$

and

$$f(u, v) = u^2 + v \sin u$$

with $h(x, y, z) = f(u(x, y, z), v(x, y, z))$.

Problem 6.16. Let (r, θ, φ) be the spherical coordinate of \mathbb{R}^3 so that

$$x = r \cos \theta \sin \varphi, \quad y = r \sin \theta \sin \varphi, \quad z = r \cos \varphi.$$

1. Find the Jacobian matrices of the map $(x, y, z) \mapsto (r, \theta, \varphi)$ and the map $(r, \theta, \varphi) \mapsto (x, y, z)$.
2. Suppose that $f(x, y, z)$ is a differential function in \mathbb{R}^3 . Express $|\nabla f|^2$ in terms of the spherical coordinate.

Problem 6.17. Let $\mathcal{U} \subseteq \mathbb{R}^n$ be open and convex, and $f : \mathcal{U} \rightarrow \mathbb{R}^m$ be differentiable on \mathcal{U} . Show that for each $a, b \in \mathcal{U}$ and vector $v \in \mathbb{R}^m$, there exists c on the line segment joining a and b such that

$$v \cdot [f(b) - f(a)] = v \cdot D(f)(c)(b - a).$$

Problem 6.18. Let $\mathcal{U} \subseteq \mathbb{R}^n$ be open, and for each $1 \leq i, j \leq n$, $a_{ij} : \mathcal{U} \rightarrow \mathbb{R}$ be differentiable functions. Define $A = [a_{ij}]$ and $J = \det(A)$. Show that

$$\frac{\partial J}{\partial x_k} = \text{tr}(\text{Adj}(A) \frac{\partial A}{\partial x_k}) \quad \forall 1 \leq k \leq n,$$

where for a square matrix $M = [m_{ij}]$, $\text{tr}(M)$ denotes the trace of M , $\text{Adj}(M)$ denotes the adjoint matrix of M , and $\frac{\partial M}{\partial x_k}$ denotes the matrix whose (i, j) -th entry is given by $\frac{\partial m_{ij}}{\partial x_k}$.

Hint: Show that

$$\frac{\partial J}{\partial x_k} = \begin{vmatrix} \frac{\partial a_{11}}{\partial x_k} & a_{12} & \cdots & a_{1n} \\ \frac{\partial a_{21}}{\partial x_k} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial a_{n1}}{\partial x_k} & a_{n2} & \cdots & a_{nn} \end{vmatrix} + \begin{vmatrix} a_{11} & \frac{\partial a_{12}}{\partial x_k} & a_{13} & \cdots & a_{1n} \\ a_{21} & \frac{\partial a_{22}}{\partial x_k} & a_{23} & \cdots & a_{2n} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ a_{n1} & \frac{\partial a_{n2}}{\partial x_k} & a_{n3} & \cdots & a_{nn} \end{vmatrix} + \cdots + \begin{vmatrix} a_{11} & \cdots & a_{(n-1)1} & \frac{\partial a_{1n}}{\partial x_k} \\ a_{21} & \cdots & a_{(n-1)2} & \frac{\partial a_{2n}}{\partial x_k} \\ \vdots & \vdots & \vdots & \vdots \\ a_{n1} & \cdots & a_{(n-1)n} & \frac{\partial a_{n1}}{\partial x_k} \end{vmatrix}$$

and rewrite this identity in the form which is asked to prove. You can also show the differentiation formula by applying the chain rule to the composite function $F \circ g$ of maps $g : \mathcal{U} \rightarrow \mathbb{R}^{n^2}$ and $F : \mathbb{R}^{n^2} \rightarrow \mathbb{R}$ defined by $g(x) = (a_{11}(x), a_{12}(x), \dots, a_{nn}(x))$ and $F(a_{11}, \dots, a_{nn}) = \det([a_{ij}])$. Check first what $\frac{\partial F}{\partial a_{ij}}$ is.

Problem 6.19. Let $\psi : \mathbb{R}^n \rightarrow \mathbb{R}^n$ be a differentiable function such that $\frac{\partial^2 \psi_k}{\partial x_i \partial x_j} = \frac{\partial}{\partial x_i} \left(\frac{\partial \psi_k}{\partial x_j} \right)$ exists and is continuous in \mathbb{R}^n for each $1 \leq i, j, k \leq n$. Suppose that $(D\psi)(x) \in \text{GL}(n)$ for all $x \in \mathbb{R}^n$, and define $A = (D\psi)^{-1}$ (or in terms of their matrix representation, $[A] = [D\psi]^{-1}$). Let $\psi = (\psi_1, \dots, \psi_n)$ and $[A] = [a_{ij}]$.

1. Show that $\sum_{k=1}^n a_{ik} \frac{\partial \psi_k}{\partial x_j} = \sum_{k=1}^n \frac{\partial \psi_i}{\partial x_k} a_{kj} = \delta_{ij}$, where δ_{ij} is the Kronecker delta; that is, $\delta_{ij} = 1$ if $i = j$ or $\delta_{ij} = 0$ if $i \neq j$.
2. Show that for each $1 \leq i, j, k \leq n$, $a_{ij} : \mathbb{R}^n \rightarrow \mathbb{R}$ is differentiable, and

$$\frac{\partial a_{ij}}{\partial x_k} = - \sum_{r,s=1}^n a_{ir} \frac{\partial^2 \psi_r}{\partial x_k \partial x_s} a_{sj}.$$

Problem 6.20. Let $\mathcal{U} \subseteq \mathbb{R}^n$ be open and connected, and $f : \mathcal{U} \rightarrow \mathbb{R}$ be a function such that $\frac{\partial f}{\partial x_j}(x) = 0$ for all $x \in \mathcal{U}$. Show that f is constant in \mathcal{U} .

§6.5 Directional Derivatives and Gradient Vectors

Problem 6.21. Let

$$f(x, y) = \begin{cases} \frac{x^3 y}{x^4 + y^2} & \text{if } (x, y) \neq (0, 0), \\ 0 & \text{if } (x, y) = (0, 0). \end{cases}$$

Show that the directional derivative of f at the origin exists in all directions u , and

$$(D_u f)(0, 0) = \left(\frac{\partial f}{\partial x}(0, 0), \frac{\partial f}{\partial y}(0, 0) \right) \cdot u.$$

§6.6 Higher Derivatives of Functions

Problem 6.22. Let $f(x, y, z) = (x^2 + 1) \cos(yz)$, and $a = (0, \frac{\pi}{2}, 1)$, $u = (1, 0, 0)$, $v = (0, 1, 0)$ and $w = (2, 0, 1)$.

1. Compute $(Df)(a)(u)$.

2. Compute $(D^2f)(a)(v)(u)$.
3. Compute $(D^3f)(a)(w)(v)(u)$.

Problem 6.23. 1. If $f : A \subseteq \mathbb{R}^n \rightarrow \mathbb{R}^m$ and $g : B \subseteq \mathbb{R}^m \rightarrow \mathbb{R}^\ell$ are twice differentiable and $f(A) \subseteq B$, then for $x_0 \in A$, $u, v \in \mathbb{R}^n$, show that

$$\begin{aligned} D^2(g \circ f)(x_0)(u, v) \\ = (D^2g)(f(x_0))((Df)(x_0)(u), Df(x_0)(v)) + (Dg)(f(x_0))((D^2f)(x_0)(u, v)). \end{aligned}$$

2. If $p : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is a linear map plus some constant; that is, $p(x) = Lx + c$ for some $L \in \mathcal{B}(\mathbb{R}^n, \mathbb{R}^m)$, and $f : A \subseteq \mathbb{R}^m \rightarrow \mathbb{R}^s$ is k -times differentiable, prove that

$$D^k(f \circ p)(x_0)(u^{(1)}, \dots, u^{(k)}) = (D^k f)(p(x_0))((Dp)(x_0)(u^{(1)}), \dots, (Dp)(x_0)(u^{(k)})).$$

§6.7 Taylor's Theorem

Problem 6.24. Let $f(x, y)$ be a real-valued function on \mathbb{R}^2 . Suppose that f is of class \mathcal{C}^1 (that is, all first partial derivatives are continuous on \mathbb{R}^2) and $\frac{\partial^2 f}{\partial x \partial y}$ exists and is continuous.

Show that $\frac{\partial^2 f}{\partial y \partial x}$ exists and $\frac{\partial^2 f}{\partial x \partial y} = \frac{\partial^2 f}{\partial y \partial x}$.

Hint: Mimic the proof of Theorem 6.81.

Problem 6.25. Let $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$ be differentiable, and Df is a constant map in $\mathcal{B}(\mathbb{R}^n, \mathbb{R}^m)$; that is, $(Df)(x_1)(u) = (Df)(x_2)(u)$ for all $x_1, x_2 \in \mathbb{R}^n$ and $u \in \mathbb{R}^n$. Show that f is a linear term plus a constant and that the linear part of f is the constant value of Df .

Problem 6.26. Let $\mathcal{U} \subseteq \mathbb{R}^n$ be open, and $f : \mathcal{U} \rightarrow \mathbb{R}^n$ be of class \mathcal{C}^2 such that $Df : \mathcal{U} \rightarrow \mathcal{B}(\mathbb{R}^n, \mathbb{R}^n)$ satisfies $(Df)(x) \in \text{GL}(n)$ for all $x \in \mathcal{U}$. Define $J = \det([Df])$ and $A = [Df]^{-1}$. With a_{ij} denoting the (i, j) -th entry of A , show the Piola identity

$$\sum_{i=1}^n \frac{\partial}{\partial x_i} (J a_{ij})(x) = 0 \quad \forall 1 \leq j \leq n \text{ and } x \in \mathcal{U}. \quad (6.9.1)$$

Is f continuous at $(0, 0)$? Is f differentiable at $(0, 0)$?

Problem 6.27. Let $\mathcal{U} \subseteq \mathbb{R}^n$ be open, and $f : \mathcal{U} \rightarrow \mathbb{R}$ be of class \mathcal{C}^k and $(D^j f)(x_0) = 0$ for $j = 1, \dots, k-1$, but $(D^k f)(x_0)(u, u, \dots, u) < 0$ for all $u \in \mathbb{R}^n$, $u \neq 0$. Show that f has a local maximum at x_0 ; that is, $\exists \delta > 0$ such that

$$f(x) \leq f(x_0) \quad \forall x \in D(x_0, \delta).$$

§6.8 Maxima and Minima

Problem 6.28. Let $f(x, y) = x^3 + x - 4xy + 2y^2$,

1. Find all critical points of f .
2. Find the corresponding quadratic form $Q(x, y, h, k)$ (or $(D^2f(x, y))((h, k), (h, k))$) at these critical points, and determine which of them is positive definite.
3. Find all relative extrema and saddle points.
4. Find the maximal value of f on the set

$$A = \{(x, y) \mid 0 \leq x \leq 1, 0 \leq y \leq 1, x + y \leq 1\}.$$

Problem 6.29. Let $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ be given by

$$f(x, y) = \begin{cases} x^2 + y^2 - 2x^2y - \frac{4x^6y^2}{(x^4 + y^2)^2} & \text{if } (x, y) \neq (0, 0), \\ 0 & \text{if } (x, y) = (0, 0). \end{cases}$$

1. Show that f is continuous (at $(0, 0)$) by showing that for all $(x, y) \in \mathbb{R}^2$,

$$4x^4y^2 \leq (x^4 + y^2)^2.$$

2. For $0 \leq \theta \leq 2\pi$, $-\infty < t < \infty$, define

$$g_\theta(t) = f(t \cos \theta, t \sin \theta).$$

Show that each g_θ has a strict local minimum at $t = 0$. In other words, the restriction of f to each straight line through $(0, 0)$ has a strict local minimum at $(0, 0)$.

3. Show that $(0, 0)$ is not a local minimum for f .

Problem 6.30 (True or False). Determine whether the following statements are true or false. If it is true, prove it. Otherwise, give a counter-example.

- 1.
- 2.

- 3.
4. Let $\mathcal{U} \subseteq \mathbb{R}^n$ be open. Then $f : \mathcal{U} \rightarrow \mathbb{R}$ is differentiable at $a \in \mathcal{U}$ if and only if each directional derivative $(D_u f)(a)$ exists and

$$(D_u f)(a) = \sum_{j=1}^n \frac{\partial f}{\partial x_j}(a) u_j = \left(\frac{\partial f}{\partial x_1}(a), \dots, \frac{\partial f}{\partial x_n}(a) \right) \cdot u$$

where $u = (u_1, \dots, u_n)$ is a unit vector.

5. Let $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ be of class \mathcal{C}^1 . Assume that all second order partial derivatives of f exist, then f is second times differentiable in \mathbb{R}^2 .
6. Let f be a function defined on \mathbb{R}^2 , and A be an invertible matrix. Define $y = Ax$ for $x \in \mathbb{R}^n$. Then $f(y)$ is differentiable if and only if $f(Ax)$ is differentiable as a function of x .
7. Let $f : [a, b] \rightarrow \mathbb{R}^2$ be continuous and be differentiable on (a, b) . If $f(a) = f(b)$, then there exists some $c \in (a, b)$ such that $f'(c) = 0$.