

EXERCISE 8.1

We have

$$\begin{aligned} f(x, y) - f(x_0, y_0) &= a(x_0 + \Delta x)^2 + 2b(x_0 + \Delta x)(y_0 + \Delta y) + c(y_0 + \Delta y)^2 - ax_0^2 - 2bx_0y_0 - cy_0^2 \\ &= (2ax_0 + 2by_0)\Delta x + (2bx_0 + 2cy_0)\Delta y + (a\Delta x^2 + 2b\Delta x\Delta y + c\Delta y^2). \end{aligned}$$

By

$$|a\Delta x^2 + 2b\Delta x\Delta y + c\Delta y^2| \leq (|a| + 2|b| + |c|) \max\{|\Delta x|, |\Delta y|\}^2 \leq (|a| + 2|b| + |c|)\epsilon \max\{|\Delta x|, |\Delta y|\},$$

the function is differentiable, with

$$f'(x_0, y_0)(u, v) = (2ax_0 + 2by_0)u + (2bx_0 + 2cy_0)v.$$

EXERCISE 8.2

Suppose $F(\vec{x})$ is differentiable at \vec{x}_0 . For any $\epsilon > 0$, there is $\delta > 0$, such that

$$\|\vec{x} - \vec{x}_0\| < \delta \implies \|F(\vec{x}) - \vec{a} - L(\Delta\vec{x})\| \leq \epsilon\|\Delta\vec{x}\|.$$

Then

$$\vec{a} = F(\vec{x}_0), \quad \|F(\vec{x}) - F(\vec{x}_0)\| \leq \|L(\Delta\vec{x})\| + \epsilon\|\Delta\vec{x}\| \leq (\|L\| + \epsilon)\|\Delta\vec{x}\| < (\|L\| + \epsilon)\delta.$$

This shows that

$$\|\Delta\vec{x}\| < \min\left\{\delta, \frac{\epsilon}{\|L\| + \epsilon}\right\} \implies \|F(\vec{x}) - F(\vec{x}_0)\| < \epsilon.$$

This proves that $F(\vec{x})$ is continuous at \vec{x}_0 .

By Exercise 6.21, the norm is always continuous. Suppose $\|\vec{x}\|_2$ has linear approximation $a + l(\vec{x})$ at $\vec{0}$. Then $a = \|\vec{0}\|_2 = 0$ and for any $\epsilon > 0$, there is $\delta > 0$, such that

$$\|\vec{x}\|_2 < \delta \implies \left| \|\vec{x}\|_2 - l(\vec{x}) \right| \leq \epsilon\|\vec{x}\|_2.$$

Applying the property to $-\vec{x}$ and using the linearity of l , we find

$$\|\vec{x}\|_2 < \delta \implies \left| \|\vec{x}\|_2 + l(\vec{x}) \right| \leq \epsilon\|\vec{x}\|_2.$$

Therefore

$$\|\vec{x}\|_2 < \delta \implies 2|l(\vec{x})| \leq \left| \|\vec{x}\|_2 - l(\vec{x}) \right| + \left| \|\vec{x}\|_2 + l(\vec{x}) \right| \leq 2\epsilon\|\vec{x}\|_2.$$

This further implies $|l(c\vec{x})| \leq \epsilon\|c\vec{x}\|_2$, where $c\vec{x}$ can be any vector of arbitrary length. Since ϵ is arbitrary, we conclude that $l(\vec{x}) = 0$ for all \vec{x} . Thus for any $\epsilon > 0$, there is $\delta > 0$, such that $\|\vec{x}\|_2 < \delta$ implies $\left| \|\vec{x}\|_2 \right| \leq \epsilon\|\vec{x}\|_2$. This cannot hold for $\epsilon < 1$. The contradiction shows that $\|\vec{x}\|_2$ is not differentiable at $\vec{0}$.

EXERCISE 8.3

Denote $L = F'(\vec{x}_0)$. For any $\epsilon > 0$, there is $\delta > 0$, such that

$$\|\vec{x} - \vec{x}_0\| < \delta \implies \|F(\vec{x}) - F(\vec{x}_0) - L(\Delta x)\| = \|F(\vec{x}) - L(\Delta x)\| \leq \epsilon \|\Delta x\|, \quad |\lambda(\vec{x}) - \lambda(\vec{x}_0)| < \epsilon.$$

Then

$$\begin{aligned} \|F(\vec{x})\| &\leq \|L(\Delta x)\| + \epsilon \|\Delta x\| \leq (\|L\| + \epsilon) \|\Delta x\|, \\ \|\lambda(\vec{x})F(\vec{x}) - \lambda(\vec{x}_0)L(\Delta x)\| &\leq \|\lambda(\vec{x})F(\vec{x}) - \lambda(\vec{x}_0)F(\vec{x})\| + \|\lambda(\vec{x}_0)F(\vec{x}) - \lambda(\vec{x}_0)L(\Delta x)\| \\ &\leq |\lambda(\vec{x}) - \lambda(\vec{x}_0)| \|F(\vec{x})\| + |\lambda(\vec{x}_0)| \|F(\vec{x}) - L(\Delta x)\| \\ &\leq \epsilon (\|L\| + \epsilon) \|\Delta x\| + \epsilon |\lambda(\vec{x}_0)| \|\Delta x\| \\ &= \epsilon (\|L\| + \epsilon + |\lambda(\vec{x}_0)|) \|\Delta x\|. \end{aligned}$$

This proves that $\lambda(\vec{x})F(\vec{x})$ is differentiable with $(\lambda F)'(\vec{x}_0) = \lambda(\vec{x}_0)L = \lambda(\vec{x}_0)F'(\vec{x}_0)$.

EXERCISE 8.4

If $f(\vec{x}) = f(\vec{x}_0) + J(\vec{x}) \cdot \Delta \vec{x}$ and J is continuous, then

$$f(\vec{x}) = f(\vec{x}_0) + J(\vec{x}_0) \cdot \Delta \vec{x} + R(\vec{x})$$

with

$$|R(\vec{x})| = |(J(\vec{x}) - J(\vec{x}_0)) \cdot \Delta \vec{x}| \leq \|J(\vec{x}) - J(\vec{x}_0)\|_2 \|\Delta \vec{x}\|_2.$$

The continuity of J at \vec{x}_0 then tells us $R(\vec{x}) = o(\|\Delta \vec{x}\|)$. Since $J(\vec{x}_0) \cdot \Delta \vec{x}$ is linear in $\Delta \vec{x}$, we conclude that f is differentiable at \vec{x}_0 and $\nabla f(\vec{x}_0) = J(\vec{x}_0)$.

Conversely, suppose f is differentiable at \vec{x}_0 . Let

$$R(\vec{x}) = f(\vec{x}) - f(\vec{x}_0) - f'(\vec{x}_0)(\Delta \vec{x}) = f(\vec{x}) - f(\vec{x}_0) - \nabla f(\vec{x}_0) \cdot \Delta \vec{x}.$$

Then

$$\lim_{\Delta \vec{x} \rightarrow \vec{0}} \left\| \frac{R(\vec{x}) \Delta \vec{x}}{\|\Delta \vec{x}\|_2^2} \right\| = \lim_{\Delta \vec{x} \rightarrow \vec{0}} \frac{|R(\vec{x})|}{\|\Delta \vec{x}\|_2} = 0.$$

Thus $J(\vec{x}) = \nabla f(\vec{x}_0) + \frac{R(\vec{x}) \Delta \vec{x}}{\|\Delta \vec{x}\|_2^2}$ is continuous at \vec{x}_0 , and

$$\begin{aligned} f(\vec{x}_0) + J(\vec{x}) \cdot \Delta \vec{x} &= f(\vec{x}_0) + \nabla f(\vec{x}_0) \cdot \Delta \vec{x} + \frac{R(\vec{x}) \Delta \vec{x}}{\|\Delta \vec{x}\|_2^2} \cdot \Delta \vec{x} \\ &= f(\vec{x}_0) + \nabla f(\vec{x}_0) \cdot \Delta \vec{x} + R(\vec{x}) = f(\vec{x}). \end{aligned}$$

EXERCISE 8.5

Suppose $f(c\vec{x}) = c^p f(\vec{x})$ for $c > 0$. If f is differentiable at $\vec{0}$. Then by restricting the definition to the straight line $t\vec{x}$ for fixed \vec{x} and changing t , $f(t\vec{x})$ is differentiable at $t = 0$. This implies $f(t\vec{x}) = t^p f(\vec{x})$ is differentiable at $t = 0^+$. Thus we conclude that either $f(\vec{x}) = 0$ for all \vec{x} or $p \geq 1$. The continuity at $\vec{0}$ then implies that $f(\vec{0}) = \vec{0}$.

Suppose $p > 1$. Then by $\vec{x} = r\vec{u}$, $r = \|\vec{x}\|$, $\|\vec{u}\| = 1$, we have $f(\vec{x}) = r^p f(\vec{u})$. If $|f(\vec{u})| < B$ is bounded on the unit sphere (i.e., for $\|\vec{u}\| = 1$), then $\|\vec{x}\| < \delta$ implies that $|f(\vec{x})| \leq \delta^p B$, $\|\vec{x}\|$,

so that F is differentiable at $\vec{0}$ with $f'(\vec{0}) = 0$. If $f(\vec{u})$ is unbounded on the unit sphere, then no matter how small r is, we can always find \vec{u} such that $|f(\vec{x})|$ is arbitrarily large. Thus the function is not continuous at $\vec{0}$. Consequently, the function is not differentiable at $\vec{0}$.

Suppose $p = 1$ and $f'(\vec{0}) = l$. Then $g(\vec{x}) = f(\vec{x}) - l(\vec{x})$ satisfies $g(c\vec{x}) = cg(\vec{x})$ for $c > 0$ and $g'(\vec{x}_0) = 0$. Thus for any $\epsilon > 0$, there is $\delta > 0$, such that $\|\vec{x}\| < \delta$ implies $|g(\vec{x})| \leq \epsilon\|\vec{x}\|$. This further implies that $|g(c\vec{x})| = |cg(\vec{x})| \leq \epsilon\|c\vec{x}\|$ for any $c > 0$. Since $c\vec{x}$ can be any nonzero vector of arbitrary length, we conclude that $|g(\vec{x})| \leq \epsilon\|\vec{x}\|$ for all \vec{x} . Since ϵ is arbitrary, we have $g(\vec{x}) = 0$ for all \vec{x} . Thus $f(\vec{x}) = l(\vec{x})$ is a linear functional.

EXERCISE 8.6

We have

$$A\vec{x} \cdot \vec{x} - A\vec{x}_0 \cdot \vec{x}_0 = A\Delta x \cdot \vec{x}_0 + A\vec{x}_0 \cdot \Delta x + A\Delta x \cdot \Delta x.$$

Since $A\Delta x \cdot \vec{x}_0 + A\vec{x}_0 \cdot \Delta x = (A + A^T)\vec{x}_0 \cdot \Delta x$ is a linear functional of Δx and $\|A\Delta x \cdot \Delta x\|_2 \leq \|A\|\|\Delta x\|_2^2 \leq \epsilon\|A\|\|\Delta x\|_2$, we conclude that $A\vec{x} \cdot \vec{x}$ is differentiable, with the derivative at \vec{x}_0 to be $(A + A^T)\vec{x}_0 \cdot \vec{v}$.

EXERCISE 8.7

We have

$$B(\vec{x}_0 + \Delta\vec{x}, \vec{y}_0 + \Delta\vec{y}) = B(\vec{x}_0, \vec{y}_0) + B(\Delta\vec{x}, \vec{y}_0) + B(\vec{x}_0, \Delta\vec{y}) + B(\Delta\vec{x}, \Delta\vec{y})$$

The term $B(\vec{x}_0, \vec{y}_0)$ is constant. The terms $B(\Delta\vec{x}, \vec{y}_0)$ and $B(\vec{x}_0, \Delta\vec{y})$ are linear in $(\Delta\vec{x}, \Delta\vec{y})$. The term $B(\Delta\vec{x}, \Delta\vec{y})$ satisfies

$$\|B(\Delta\vec{x}, \Delta\vec{y})\| \leq \|B\|\|\Delta\vec{x}\|\|\Delta\vec{y}\| \leq \|B\|\|(\Delta\vec{x}, \Delta\vec{y})\|^2,$$

where $\|(\Delta\vec{x}, \Delta\vec{y})\| = \max\{\|\Delta\vec{x}\|, \|\Delta\vec{y}\|\}$. Therefore the first order derivative is given by the two linear terms

$$B'(\vec{x}_0, \vec{y}_0)(\Delta\vec{x}, \Delta\vec{y}) = B(\Delta\vec{x}, \vec{y}_0) + B(\vec{x}_0, \Delta\vec{y}).$$

EXERCISE 8.8

Let

$$\begin{aligned} \phi(t) &= \phi(t_0) + \phi'(t_0)\Delta t + R_1, & R_1 &= o(\Delta t); \\ \psi(t) &= \psi(t_0) + \psi'(t_0)\Delta t + R_2, & R_2 &= o(\Delta t). \end{aligned}$$

Then

$$\begin{aligned} B(\phi(t), \psi(t)) &= B(\phi(t_0) + \phi'(t_0)\Delta t + R_1, \psi(t_0) + \psi'(t_0)\Delta t + R_2) \\ &= B(\phi(t_0), \psi(t_0)) + [B(\phi'(t_0), \psi(t_0)) + B(\phi(t_0), \psi'(t_0))]\Delta t + R, \\ R &= B(\phi(t_0), R_2) + B(R_1, \psi(t_0)) + [B(\phi'(t_0), R_2) + B(R_1, \psi'(t_0))]\Delta t \\ &\quad + B(\phi'(t_0), \psi'(t_0))\Delta t^2 + B(R_1, R_2). \end{aligned}$$

By $\|B(\vec{u}, \vec{v})\| \leq \|B\|\|\vec{u}\|\|\vec{v}\|$, we have

$$\|B(\phi(t_0), R_2)\| \leq \|B\|\|\phi(t_0)\|\|R_2\|, \quad \|B(\phi'(t_0), R_2)\| \leq \|B\|\|\phi'(t_0)\|\|R_2\|,$$

$$\|B(\phi'(t_0), \psi'(t_0))\| \leq \|B\| \|\phi'(t_0)\| \|\psi'(t_0)\|, \quad \|B(R_1, R_2)\| \leq \|B\| \|R_1\| \|R_2\|.$$

Then by $R_1 = o(\Delta t)$ and $R_2 = o(\Delta t)$, we get $B(\phi(t_0), R_2) = o(\Delta t)$, $B(\phi'(t_0), R_2) = o(\Delta t)$, $B(\phi'(t_0), \psi'(t_0))\Delta t^2 = o(\Delta t)$, $B(R_1, R_2) = o(\Delta t)$ and similar estimations for the other terms in R . We conclude that $R = o(\Delta t)$, so that $B(\phi(t_0), \psi(t_0)) + [B(\phi'(t_0), \psi(t_0)) + B(\phi(t_0), \psi'(t_0))]\Delta t$ is linear approximation of $B(\phi(t), \psi(t))$ at t_0 . In particular, we get

$$[B(\phi(t), \psi(t))]'(t_0) = B(\phi'(t_0), \psi(t_0)) + B(\phi(t_0), \psi'(t_0)).$$

EXERCISE 8.9

By b linear in the first variable, we have

$$\left. \frac{d}{dt} \right|_{t=t_0} b(\phi(t), \vec{w}) = \lim_{t \rightarrow t_0} \frac{b(\phi(t), \vec{w}) - b(\phi(t_0), \vec{w})}{t - t_0} = \lim_{t \rightarrow t_0} b \left(\frac{\phi(t) - \phi(t_0)}{t - t_0}, \vec{w} \right).$$

Since b is a dual pairing, by Exercise 7.35, the right side equals $b(\vec{v}, \vec{w})$ for all \vec{w} if and only if

$$\phi'(t_0) = \lim_{t \rightarrow t_0} \frac{\phi(t) - \phi(t_0)}{t - t_0} = \vec{v}.$$

EXERCISE 8.10

Let $H = X - A$ (this is ΔX). Then $F(X) - F(A) = (A + H)^T(A + H) - A^T A = A^T H + H^T A + H^T H$. Since $A^T H + H^T A$ is linear in H and $\|H^T H\| \leq \|H^T\| \|H\| \leq c \|H\|^2 \leq c\epsilon \|H\|$ ($\|H^T\|$ is also a norm of H , so that $\|H^T\| \leq c \|H\|$). Thus F is differentiable, with $F'(A)(H) = A^T H + H^T A$.

EXERCISE 8.11

Suppose $f(X)$ is function of matrix that is multilinear with respect to the columns of X . Then we have

$$\begin{aligned} f(A + H) &= f(\vec{a}_1 + \vec{h}_1, \vec{a}_2 + \vec{h}_2, \dots, \vec{a}_n + \vec{h}_n) \\ &= f(\vec{a}_1, \vec{a}_2, \dots, \vec{a}_n) \\ &\quad + \sum_{1 \leq i \leq n} f(\vec{a}_1, \vec{a}_2, \dots, \vec{h}_i, \dots, \vec{a}_n) \\ &\quad + \sum_{1 \leq i < j \leq n} f(\vec{a}_1, \vec{a}_2, \dots, \vec{h}_i, \dots, \vec{h}_j, \dots, \vec{a}_n) \\ &\quad + \dots \\ &\quad + f(\vec{h}_1, \vec{h}_2, \dots, \vec{h}_n). \end{aligned}$$

The first order part gives us the derivative of f at A

$$f'(A)(H) = \sum_{1 \leq i \leq n} f(\vec{a}_1, \vec{a}_2, \dots, \vec{h}_i, \dots, \vec{a}_n).$$

In case $f = \det$ and the rank of $\vec{a}_1, \vec{a}_2, \dots, \vec{a}_n$ is $\leq n - 2$, replacing one vector gives $\vec{a}_1, \vec{a}_2, \dots, \vec{h}_i, \dots, \vec{a}_n$, with rank at most $n - 1$. Therefore all the determinants vanish, and $\det'(A)(H) = 0$.

EXERCISE 8.12

We have

$$(A + H)^k - A^k = A^{k-1}H + A^{k-2}HA + \dots + AHA^{k-2} + AH^{k-1} + R,$$

where $A^{k-1}H + A^{k-2}HA + \dots + AHA^{k-2} + AH^{k-1}$ is linear in H , and R is a finite sum of products of i copies of H and $k - i$ copies of A in all possible order, with $i \geq 2$. Because $i \geq 2$, the norm of each such product is $\leq \|A\|^{k-i}\|H\|^i \leq \|A\|^{k-i}\epsilon\|H\|$. Thus $A^{k-1}H + A^{k-2}HA + \dots + AHA^{k-2} + AH^{k-1}$ is the derivative of X^k at A .

EXERCISE 8.13

For $F(X) = X^{-1}$, we have

$$\begin{aligned} F(I + H) &= I - H + H^2 - H^3 + \dots \\ &= I - H + H^2(I - H + H^2 - H^3 + \dots) \\ &= F(I) - H + H^2(I + H)^{-1}. \end{aligned}$$

By Exercise 7.17, for $\|H\| < 1$ we also have

$$\|(I + H)^{-1} - I\| \leq \frac{\|H\|\|I\|^2}{1 - \|H\|\|I\|} = \frac{\|H\|}{1 - \|H\|},$$

so that

$$\|(I + H)^{-1}\| \leq \|I\| + \frac{\|H\|}{1 - \|H\|} = \frac{1}{1 - \|H\|}.$$

Thus

$$\|H^2(I + H)^{-1}\| \leq \|H\|^2\|(I + H)^{-1}\| \leq \frac{\|H\|^2}{1 - \|H\|} \leq \epsilon\|H\|.$$

This shows that the linear map $-H$ of H is the derivative of $F(X)$ at I .

EXERCISE 8.14 (1)

We have $f(x, 0) = |x|^{2p} \sin \frac{1}{x^2}$. Thus $f_x(0, 0)$ exists if and only if $2p > 1$. Conversely, for $p > \frac{1}{2}$, we have $f_x(0, 0) = f_y(0, 0) = 0$.

If the function is differentiable at $(0, 0)$, then $p > \frac{1}{2}$ and the linear approximation is 0. This means that for any $\epsilon > 0$, there is $\delta > 0$, such that $\sqrt{x^2 + y^2} < \delta$ implies $\left| (x^2 + y^2)^p \sin \frac{1}{(x^2 + y^2)} \right| < \epsilon\sqrt{x^2 + y^2}$. Letting $r = \sqrt{x^2 + y^2}$, we find $0 < r < \delta$ implies $\left| r^{2p} \sin \frac{1}{r^2} \right| \leq \epsilon r$. Since $p > \frac{1}{2}$, this always happens. Thus f is differentiable at $(0, 0)$ if and only if $p > \frac{1}{2}$.

EXERCISE 8.14 (2)

The function is continuous at $(0, 0)$ if and only if $p > 0$ or $q > 0$.

When $p > 0$, we have $f(0, y) = 0$ for all y . The partial derivative $f_y(0, 0) = 0$ exists. When $p = 0$, we have $f(0, y) = \begin{cases} |y|^q \sin \frac{1}{y^2} & \text{if } y \neq 0 \\ 0 & \text{if } y = 0 \end{cases}$, and $f_y(0, y)$ exists (and must be zero) if and only if $q > 1$. Similarly, $f_x(0, 0)$ exists if and only if $q > 0$ or $p > 1, q = 0$.

Suppose $p > 0$ and $q > 0$. If the function is differentiable at $(0, 0)$, then by the computation of the partial derivatives, the linear approximation must be 0. Therefore we need $\lim_{(x,y) \rightarrow (0,0)} \frac{|x|^p |y|^q}{\max\{|x|, |y|\}} \sin \frac{1}{x^2 + y^2} = 0$. By restricting the limit to the line $|x| = |y|$, we get $p + q > 1$.

Now assume $p > 0, q > 0$ and $p + q > 1$, then $|f(x, y)| \leq \|(x, y)\|_\infty^{p+q} \leq \epsilon \|(x, y)\|_\infty$ when $\|(x, y)\|_\infty < \epsilon^{\frac{1}{p+q-1}}$. Therefore the function is differentiable at $(0, 0)$ with 0 as the linear approximation.

EXERCISE 8.14 (3)

By $f(x, 0) = f(0, y) = 0$, the function has partial derivatives $f_x(0, 0) = f_y(0, 0) = 0$.

If the function is differentiable at $(0, 0)$, then the linear approximation is 0. This means that

$$\lim_{x,y \rightarrow 0^+} \frac{x^p y^q}{(x^m + y^n)^k (x + y)} = 0.$$

When restricted to $x^m = y^n$, we have

$$\frac{x^{p+q\frac{m}{n}}}{2^{k+1} x^{mk} x^{\frac{1}{n} \min\{m,n\}}} \leq \frac{x^p y^q}{(x^m + y^n)^k (x + y)} = \frac{x^{p+q\frac{m}{n}}}{2^k x^{mk} (x + x^{\frac{m}{n}})} \leq \frac{x^{p+q\frac{m}{n}}}{2^k x^{mk} x^{\frac{1}{n} \min\{m,n\}}}.$$

Therefore the restriction converges to 0 if and only if $p + q\frac{m}{n} > mk + \frac{1}{n} \min\{m, n\}$. This is the same as

$$\frac{p}{m} + \frac{q}{n} > k + \frac{\min\{m, n\}}{mn}.$$

When restricted to $x = y$, we have

$$\frac{x^{p+q}}{2^{k+1} x^{\min\{m,n\}k} x} \leq \frac{x^p y^q}{(x^m + y^n)^k (x + y)} = \frac{x^{p+q}}{2(x^m + x^n)^k x} \leq \frac{x^{p+q}}{2x^{\min\{m,n\}k} x}.$$

Therefore the restriction converges to 0 if and only if

$$p + q > \min\{m, n\}k + 1.$$

Conversely, assume the two inequalities hold. Without loss of generality, assume $m \geq n$. Then the two inequalities mean that

$$\frac{p}{m} + \frac{q}{n} > k + \frac{1}{m}, \quad p + q > nk + 1.$$

Then we consider three regions.

For $y \leq x^{\frac{m}{n}}$, we have

$$\frac{x^p y^q}{(x^m + y^n)^k (x + y)} \leq \frac{x^{p+q\frac{m}{n}}}{x^{mk} x} = x^{p+q\frac{m}{n}-mk-1}.$$

By $p + q\frac{m}{n} - mk - 1 = m \left(\frac{p}{m} + \frac{q}{n} - k - \frac{1}{m} \right) > 0$, the restriction of the limit on the region converges to 0.

For $x^{\frac{m}{n}} \leq y \leq x$, we have

$$\frac{x^p y^q}{2^k y^{nk} 2x} \leq \frac{x^p y^q}{(x^m + y^n)^k (x + y)} \leq \frac{x^p y^q}{y^{nk} x}.$$

Therefore $\lim_{x^{\frac{m}{n}} \leq y \leq x; x, y \rightarrow 0^+} \frac{x^p y^q}{(x^m + y^n)^k (x + y)} = 0$ if and only if $\lim_{x^{\frac{m}{n}} \leq y \leq x; x, y \rightarrow 0^+} x^{p-1} y^{q-nk} = 0$. For the given x , the maximum and the minimum of $x^{p-1} y^{q-nk}$ for the y in the range are $x^{p-1} x^{q-nk} = x^{p+q-nk-1}$ and $x^{p-1} x^{\frac{m}{n}(q-nk)} = x^{m(\frac{p}{m} + \frac{q}{n} - k - \frac{1}{m})}$. Therefore $\lim_{x^{\frac{m}{n}} \leq y \leq x; x, y \rightarrow 0^+} x^{p-1} y^{q-nk} = 0$ if and only if $\lim_{x \rightarrow 0^+} x^{p+q-nk-1} = 0$ and $\lim_{x \rightarrow 0^+} x^{m(\frac{p}{m} + \frac{q}{n} - k - \frac{1}{m})} = 0$. This is true, given the two inequalities.

For $y \geq x$, we have

$$\frac{x^p y^q}{(x^m + y^n)^k (x + y)} \leq \frac{y^{p+q}}{y^{nk} y} = y^{p+q-nk-1}.$$

Since $p + q > nk + 1$, the restriction of the limit on the region converges to 0.

We conclude that the function is differentiable at $(0, 0)$ if and only if

$$\frac{p}{m} + \frac{q}{n} > k + \frac{\min\{m, n\}}{mn}, \quad p + q > \min\{m, n\}k + 1.$$

EXERCISE 8.14 (4)

By $f(x, 0) = x^{pr-mk}$ and $f(0, y) = y^{qr-nk}$, the function has partial derivatives if and only if $pr - mk > 1$ and $qr - nk > 1$. Moreover, we have $f_x(0, 0) = f_y(0, 0) = 0$ when the condition is satisfied.

If the function is differentiable at $(0, 0)$, then $pr - mk > 1$ and $qr - nk > 1$, the linear approximation is 0, and we have

$$\lim_{x, y \rightarrow 0^+} \frac{(x^p + y^q)^r}{(x^m + y^n)^k (x + y)} = 0.$$

Denote

$$\lambda = \min \left\{ \frac{p}{m}, \frac{q}{n} \right\}, \quad \mu = \min\{m, n\}, \quad \nu = \min\{p, q\}.$$

As in Exercise 6.23(2), restricting the limit to $x^m = y^n$, we have

$$x^{\lambda mr} \leq (x^p + y^q)^r \leq 2^r x^{\lambda mr}, \quad (x^m + y^n)^k = 2^k x^{mk}, \quad x^{\frac{\mu}{n}} \leq x + y \leq 2x^{\frac{\mu}{n}}.$$

Therefore the restriction has limit 0 if and only if $\lim_{x \rightarrow 0^+} \frac{x^{\lambda mr}}{x^{mk} x^{\frac{\mu}{n}}} = 0$. This means that $\lambda mr > mk + \frac{\mu}{n}$, which is the same as

$$r \min \left\{ \frac{p}{m}, \frac{q}{n} \right\} > k + \frac{\min\{m, n\}}{mn}.$$

On the other hand, restricting the limit to $x = y$, we have

$$x^{\nu r} \leq (x^p + y^q)^r \leq 2^r x^{\nu r}, \quad x^{\mu k} \leq (x^m + y^n)^k \leq 2^k x^{\mu k}, \quad x + y = 2x.$$

Therefore the restriction has limit 0 if and only if $\lim_{x \rightarrow 0^+} \frac{x^{\nu r}}{x^{\mu k} x} = 0$. This means that $\nu r > \mu k + 1$, which is the same as

$$r \min\{p, q\} > k \min\{m, n\} + 1.$$

Conversely, assume the two inequalities hold. Without loss of generality, assume $m \geq n$. Then the two inequalities mean that

$$r\lambda = r \min \left\{ \frac{p}{m}, \frac{q}{n} \right\} > k + \frac{1}{m}, \quad r\nu = r \min\{p, q\} > kn + 1.$$

Then we consider three regions.

For $y \leq x^{\frac{m}{n}}$, we have

$$\frac{(x^p + y^q)^r}{(x^m + y^n)^k (x + y)} \leq \frac{(x^p + x^{q\frac{m}{n}})^r}{x^{mk} x} \leq \frac{2^r x^{\lambda mr}}{x^{mk} x}.$$

Since $\lambda mr - mk - 1 = m \left(r\lambda - k - \frac{1}{m} \right) > 0$, the restriction of the limit on the region converges to 0.

For $y \geq x$, we have

$$\frac{(x^p + y^q)^r}{(x^m + y^n)^k (x + y)} \leq \frac{(y^p + y^q)^r}{y^{nk} y} \leq \frac{2^r y^{\nu r}}{y^{nk} y}.$$

Since $\nu r - nk - 1 > 0$, the restriction of the limit on the region converges to 0.

For $x^{\frac{m}{n}} \leq y \leq x$, we have

$$y^{nk} x \leq (x^m + y^n)^k (x + y) \leq 2^k y^{nk} 2x.$$

Moreover, we have

$$\frac{1}{2}(u^r + v^r) \leq \max\{u^r, v^r\} \leq (u + v)^r \leq (2 \max\{u, v\})^r = 2^r (\max\{u, v\})^r \leq 2^r (u^r + v^r).$$

By substituting $u = x^p$ and $v = y^q$, we see that the restriction of the limit to the region converges to 0 if and only if the limit of $\frac{x^{pr} + y^{qr}}{y^{nk} x}$ converges to 0, which is the same as the

limits of $\frac{x^{pr}}{y^{nk}x}$ and $\frac{y^{qr}}{y^{nk}x}$ converge to 0. For fixed x , the maxima and minima of $\frac{x^{pr}}{y^{nk}x}$ and $\frac{y^{qr}}{y^{nk}x}$ for the y in the range are $x^{pr-nk-1}$, $x^{pr-mk-1}$, $x^{qr-nk-1}$, $x^{qr\frac{m}{n}-mk-1}$. Therefore the limit converges to zero if and only if

$$pr - nk - 1 > 0, \quad pr - mk - 1 > 0, \quad qr - nk - 1 > 0, \quad qr\frac{m}{n} - mk - 1 > 0.$$

Since the two inequalities imply all four inequalities above, we conclude that the two inequalities are necessary and sufficient.

In conclusion, the function is differentiable at $(0, 0)$ if and only if

$$r \min \left\{ \frac{p}{m}, \frac{q}{n} \right\} > k + \frac{\min\{m, n\}}{mn}, \quad r \min\{p, q\} > k \min\{m, n\} + 1.$$

EXERCISE 8.15

In Example 8.1.3, we get

$$\|\vec{x}_0 + \Delta\vec{x}\|_2^2 = \|\vec{x}_0\|_2^2 + 2\vec{x}_0 \cdot \Delta\vec{x} + \|\Delta\vec{x}\|_2^2.$$

So the gradient should give the linear functional $2\vec{x}_0 \cdot \Delta\vec{x}$. This means that $\nabla(\vec{x} \cdot \vec{x}) = 2\vec{x}$.

EXERCISE 8.16

By the computation in Exercise 8.11, for 2×2 matrix, we have

$$\begin{aligned} \det'(A)(H) &= \det(\vec{a}_1, \vec{h}_2) + \det(\vec{h}_1, \vec{a}_2) \\ &= a_{11}h_{22} - a_{21}h_{12} + a_{22}h_{11} - a_{12}h_{21} \\ &= \begin{pmatrix} a_{22} & -a_{21} \\ -a_{12} & a_{11} \end{pmatrix} \cdot \begin{pmatrix} h_{11} & h_{12} \\ h_{21} & h_{22} \end{pmatrix}. \end{aligned}$$

Here \cdot is the dot product in \mathbb{R}^4 . Therefore the gradient

$$\nabla \det = \begin{pmatrix} a_{22} & -a_{21} \\ -a_{12} & a_{11} \end{pmatrix}.$$

EXERCISE 8.17

For $X = \begin{pmatrix} x & y \\ z & w \end{pmatrix}$, we have $X^2 = \begin{pmatrix} x^2 + yz & xy + yw \\ xz + zw & yz + w^2 \end{pmatrix}$. If we identify the 2×2 matrix X with $(x, y, z, w)^T \in \mathbb{R}^4$, then the Jacobian matrix is

$$(X^2)' = \begin{pmatrix} \frac{\partial(x^2 + yz)}{\partial x} & \frac{\partial(x^2 + yz)}{\partial y} & \frac{\partial(x^2 + yz)}{\partial z} & \frac{\partial(x^2 + yz)}{\partial w} \\ \frac{\partial(xy + yw)}{\partial x} & \frac{\partial(xy + yw)}{\partial y} & \frac{\partial(xy + yw)}{\partial z} & \frac{\partial(xy + yw)}{\partial w} \\ \frac{\partial(xz + zw)}{\partial x} & \frac{\partial(xz + zw)}{\partial y} & \frac{\partial(xz + zw)}{\partial z} & \frac{\partial(xz + zw)}{\partial w} \\ \frac{\partial(yz + w^2)}{\partial x} & \frac{\partial(yz + w^2)}{\partial y} & \frac{\partial(yz + w^2)}{\partial z} & \frac{\partial(yz + w^2)}{\partial w} \end{pmatrix} = \begin{pmatrix} 2x & z & y & 0 \\ y & x + w & 0 & y \\ z & 0 & x + w & z \\ 0 & z & y & 2w \end{pmatrix}.$$

Applying the matrix to the matrix $H = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$, which corresponds to the vector $(a, b, c, d)^T \in \mathbb{R}^4$, we get

$$(X^2)'(H) = \begin{pmatrix} 2x & z & y & 0 \\ y & x+w & 0 & y \\ z & 0 & x+w & z \\ 0 & z & y & 2w \end{pmatrix} \begin{pmatrix} a \\ b \\ c \\ d \end{pmatrix} = \begin{pmatrix} 2xa + zb + yc \\ ya + (x+w)b + yd \\ za + (x+w)c + zd \\ zb + yc + 2wd \end{pmatrix}.$$

Translated back into matrix, this means

$$\begin{aligned} (X^2)'(H) &= \begin{pmatrix} 2xa + zb + yc & ya + (x+w)b + yd \\ za + (x+w)c + zd & zb + yc + 2wd \end{pmatrix} \\ &= \begin{pmatrix} x & y \\ z & w \end{pmatrix} \begin{pmatrix} a & b \\ c & d \end{pmatrix} + \begin{pmatrix} a & b \\ c & d \end{pmatrix} \begin{pmatrix} x & y \\ z & w \end{pmatrix} = XH + HX. \end{aligned}$$

This recovers the computation in Example 8.1.5.

EXERCISE 8.18

For $X = \begin{pmatrix} x & y \\ z & w \end{pmatrix}$, we have $X^{-1} = \frac{1}{xw - yz} \begin{pmatrix} w & -y \\ -z & x \end{pmatrix}$. If we identify the 2×2 matrix X with $(x, y, z, w)^T \in \mathbb{R}^4$, then the Jacobian matrix is

$$\begin{aligned} (X^{-1})' &= \frac{\partial}{\partial X} \left[\frac{1}{xw - yz} \begin{pmatrix} w \\ -y \\ -z \\ x \end{pmatrix} \right] = \frac{\partial}{\partial X} \begin{pmatrix} w \\ -y \\ -z \\ x \end{pmatrix} + \begin{pmatrix} w \\ -y \\ -z \\ x \end{pmatrix} \frac{\partial}{\partial X} \left[\frac{1}{xw - yz} \right] \\ &= \begin{pmatrix} 0 & 0 & 0 & 1 \\ 0 & -1 & 0 & 0 \\ 0 & 0 & -1 & 0 \\ 1 & 0 & 0 & 0 \end{pmatrix} - \frac{1}{(xw - yz)^2} \begin{pmatrix} w \\ -y \\ -z \\ x \end{pmatrix} (w - z - yx) \\ &= \begin{pmatrix} 0 & 0 & 0 & 1 \\ 0 & -1 & 0 & 0 \\ 0 & 0 & -1 & 0 \\ 1 & 0 & 0 & 0 \end{pmatrix} - \frac{1}{(xw - yz)^2} \begin{pmatrix} w^2 & -zw & -yw & xw \\ -yw & yz & y^2 & -xy \\ -zw & z^2 & yz & -xz \\ xw & -xz & -xy & x^2 \end{pmatrix} \end{aligned}$$

Then at the identity matrix, we have

$$(X^{-1})'_{\text{at } I} = \begin{pmatrix} 0 & 0 & 0 & 1 \\ 0 & -1 & 0 & 0 \\ 0 & 0 & -1 & 0 \\ 1 & 0 & 0 & 0 \end{pmatrix} - \frac{1}{1^2} \begin{pmatrix} 1 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 1 \end{pmatrix} = \begin{pmatrix} -1 & 0 & 0 & 0 \\ 0 & -1 & 0 & 0 \\ 0 & 0 & -1 & 0 \\ 0 & 0 & 0 & -1 \end{pmatrix} = -\text{identity}.$$

This recovers the computation in Exercise 8.13.

EXERCISE 8.19

Let $\nabla f = (a, b, c)$. Then the condition tells us

$$\begin{aligned} 1 &= \nabla f \cdot \frac{(1, 2, 2)}{\|(1, 2, 2)\|} = \frac{1}{3}(a + 2b + 2c), \\ \sqrt{2} &= \nabla f \cdot \frac{(0, 1, -1)}{\|(0, 1, -1)\|} = \frac{1}{\sqrt{2}}(b - c), \\ 3 &= \nabla f \cdot \frac{(0, 0, 1)}{\|(0, 0, 1)\|} = c. \end{aligned}$$

Solving the system, we get $\nabla f = (-15, 5, 3)$.

EXERCISE 8.20

Let $\nabla f = a_1\vec{u}_1 + a_2\vec{u}_2 + \cdots + a_n\vec{u}_n$. Then by the orthonormal property, we have $D_{\vec{u}_1}f = \nabla f \cdot \vec{u}_1 = a_1\vec{u}_1 \cdot \vec{u}_1 + a_2\vec{u}_2 \cdot \vec{u}_1 + \cdots + a_n\vec{u}_n \cdot \vec{u}_1 = a_11 + a_20 + \cdots + a_n0 = a_1$. The other coefficients can be similarly derived.

EXERCISE 8.21 (1)

$$r_x = \frac{x}{\sqrt{x^2 + y^2}}, r_y = \frac{y}{\sqrt{x^2 + y^2}}, \theta_x = \frac{1}{1 + \frac{y^2}{x^2}} \frac{-y}{x^2} = \frac{-y}{x^2 + y^2}, \theta_y = \frac{1}{1 + \frac{y^2}{x^2}} \frac{y}{x} = \frac{x}{x^2 + y^2}.$$

$$\text{The Jacobian matrix } \frac{\partial(r, \theta)}{\partial(x, y)} = \frac{1}{x^2 + y^2} \begin{pmatrix} x\sqrt{x^2 + y^2} & y\sqrt{x^2 + y^2} \\ -y & x \end{pmatrix}.$$

$$\text{The differential } \begin{pmatrix} dr \\ d\theta \end{pmatrix} = \begin{pmatrix} \sqrt{x^2 + y^2}(xdx + ydy) \\ -ydx + xdy \end{pmatrix}.$$

EXERCISE 8.21 (2)

$$\text{The Jacobian matrix } \frac{\partial(u_1, u_2, u_3)}{\partial(x_1, x_2, x_3)} = \begin{pmatrix} 1 & 1 & 1 \\ x_2 + x_3 & x_3 + x_1 & x_1 + x_2 \\ x_2x_3 & x_3x_1 & x_1x_2 \end{pmatrix}.$$

$$\text{The differential } \begin{pmatrix} du_1 \\ du_2 \\ du_3 \end{pmatrix} = \begin{pmatrix} dx_1 + dx_2 + dx_3 \\ (x_2 + x_3)dx_1 + (x_3 + x_1)dx_2 + (x_1 + x_2)dx_3 \\ x_2x_3dx_1 + x_3x_1dx_2 + x_1x_2dx_3 \end{pmatrix}.$$

EXERCISE 8.21 (3)

$$\text{The Jacobian matrix } \frac{\partial(x, y, z)}{\partial(r, \phi, \theta)} = \begin{pmatrix} \sin \phi \cos \theta & r \cos \phi \cos \theta & -r \sin \phi \sin \theta \\ \sin \phi \sin \theta & r \cos \phi \sin \theta & r \sin \phi \cos \theta \\ \cos \phi & -r \sin \phi & 0 \end{pmatrix}.$$

$$\text{The differential } \begin{pmatrix} dx \\ dy \\ dz \end{pmatrix} = \begin{pmatrix} \sin \phi \cos \theta dr + r \cos \phi \cos \theta d\phi - r \sin \phi \sin \theta d\theta \\ \sin \phi \sin \theta dr + r \cos \phi \sin \theta d\phi + r \sin \phi \cos \theta d\theta \\ \cos \phi dr - r \sin \phi d\phi \end{pmatrix}.$$

EXERCISE 8.22

Suppose $\|F(\vec{x})\| < B$. Then $\|\vec{x}\|_2 < \delta$ implies $\|\|\vec{x}\|_2^2 F(\vec{x})\| = \|\vec{x}\|_2^2 \|F(\vec{x})\| < \delta B \|\vec{x}\|_2$. This shows that $\|\vec{x}\|_2^2 F(\vec{x})$ is differentiable at $\vec{0}$, with derivatives O .

On the other hand, if we choose $f(\vec{x}) = k$ if k coordinates are rational and $n - k$ coordinates are irrational. Then $\|\vec{x}\|_2^2 f(\vec{x})$ is not continuous along any coordinate except $\vec{0}$. Thus the function has no partial derivatives away from $\vec{0}$.

EXERCISE 8.23

The function $f(x, y) = \begin{cases} (x^2 + y^2) \sin \frac{1}{x^2 + y^2} & \text{if } (x, y) \neq (0, 0) \\ 0 & \text{if } (x, y) = (0, 0) \end{cases}$ in Exercise 8.14 is differentiable everywhere. But $f(x, 0) = x^2 \sin \frac{1}{x^2}$, and $f_x(x, 0) = 2x \sin \frac{1}{x^2} - \frac{2}{x} \cos \frac{1}{x^2}$ is not continuous at $x = 0$. The other partial derivative is also not continuous.

EXERCISE 8.24

By the existence of $f_x(x_0, y_0)$, for any $\epsilon > 0$, there is $\delta > 0$, such that $|\Delta x| < \delta$ implies

$$|f(x, y_0) - f(x_0, y_0) - f_x(x_0, y_0)\Delta x| \leq \epsilon|\Delta x|.$$

By mean value theorem, we have $f(x, y) - f(x, y_0) = f_y(x, d)\Delta y$. Then by the continuity of f_y at (x_0, y_0) , there is $\delta' > 0$, such that $|\Delta x| < \delta'$, $|\Delta y| < \delta'$ implies $|f_y(x, d) - f_y(x_0, y_0)| < \epsilon$. Therefore

$$|f(x, y) - f(x, y_0) - f_y(x_0, y_0)\Delta y| = |(f_y(x, d) - f_y(x_0, y_0))\Delta y| \leq \epsilon|\Delta y|.$$

Combining the two estimations, for $|\Delta x| < \min\{\delta, \delta'\}$, $|\Delta y| < \min\{\delta, \delta'\}$, we have

$$\begin{aligned} & |f(x, y) - f(x_0, y_0) - f_x(x_0, y_0)\Delta x - f_y(x_0, y_0)\Delta y| \\ & \leq |f(x, y_0) - f(x_0, y_0) - f_x(x_0, y_0)\Delta x| + |f(x, y) - f(x, y_0) - f_y(x_0, y_0)\Delta y| \\ & \leq \epsilon|\Delta x| + \epsilon|\Delta y|. \end{aligned}$$

This proves the differentiability of f at (x_0, y_0) .

In general, if one partial derivative exist and the other partial derivatives are continuous, then the function is differentiable.

EXERCISE 8.25

If $F(x, y)$ is differentiable, then the composition (or restriction) $F(x, x) = f'(x)$ is differentiable, which means the existence of second order derivative.

Conversely, suppose f has second order derivative. By Proposition 3.4.2 we have

$$f(x) = f(x_0) + f'(x_0)\Delta x + \frac{1}{2}f''(x_0)\Delta x^2 + R(x), \quad \lim_{x \rightarrow x_0} \frac{R(x)}{\Delta x^2} = 0.$$

We also note that $R(x)$ has second order derivative, and $R(x_0) = R'(x_0) = R''(x_0) = 0$. The Taylor expansion implies

$$\begin{aligned} f(x) - f(y) &= f'(x_0)(\Delta x - \Delta y) + \frac{1}{2}f''(x_0)(\Delta x^2 - \Delta y^2) + R(x) - R(y) \\ &= f'(x_0)(x - y) + \frac{1}{2}f''(x_0)(x - y)(\Delta x + \Delta y) + R(x) - R(y). \end{aligned}$$

Then for $x \neq y$ and (x, y) near (x_0, x_0) , we get

$$\frac{f(x) - f(y)}{x - y} = f'(x_0) + \frac{1}{2}f''(x_0)(\Delta x + \Delta y) + \frac{R(x) - R(y)}{x - y}.$$

If we can show that $\frac{R(x) - R(y)}{x - y} = o(\|(\Delta x, \Delta y)\|)$, then we find $F(x, y)$ is differentiable at (x_0, x_0) with

$$F'(x_0, x_0)(u, v) = \frac{1}{2}f''(x_0)(u + v).$$

By the mean value theorem, we have $\frac{R(x) - R(y)}{x - y} = R'(c)$, where c is between x and y . Using $R'(x_0) = R''(x_0) = 0$, we further have

$$R'(c) = R'(x_0) + R''(x_0)(c - x_0) + o(c - x_0) = o(c - x_0).$$

Since c is between x and y , we have $|c - x_0| \leq \max\{|x - x_0|, |y - x_0|\} = \|(\Delta x, \Delta y)\|_\infty$. Therefore

$$\frac{R(x) - R(y)}{x - y} = R'(c) = o(\|(\Delta x, \Delta y)\|_\infty).$$

This completes the proof that F is differentiable at (x_0, x_0) .

In case $x_0 \neq y_0$, both $f(x) - f(y)$ and $x - y$ are differentiable at (x_0, y_0) and $x - y \neq 0$ near (x_0, y_0) . Therefore the quotient $F(x, y)$ is differentiable at (x_0, y_0) .

Additional: F' is continuous if and only if f'' is continuous.

We already know that, if f has second order derivative, then

$$F_x(x, y) = \begin{cases} \frac{f'(x)(x - y) - f(x) + f(y)}{(x - y)^2}, & \text{if } x \neq y, \\ \frac{1}{2}f''(x), & \text{if } x = y, \end{cases}$$

and we have the similar formula for F_y . The continuity of F' means the continuity of F_x and F_y . The formula tells us that the continuity of $F_x(x, x)$ already means f'' is continuous.

Conversely, suppose $f''(x)$ is continuous. We want to show that F_x is continuous. The continuity of f' already implies the continuity of F_x at (x_0, y_0) whenever $x_0 \neq y_0$. It remains to show the continuity at (x_0, x_0) .

For $x \neq y$, we have

$$\begin{aligned} F_x(x, y) - \frac{1}{2}f''(x_0) &= \frac{f(y) - f(x) - f'(x)(y - x) - \frac{1}{2}f''(x_0)(y - x)^2}{(y - x)^2} \\ &= \frac{r_1(y, x)}{(y - x)^2} = \frac{r_1(y, x) - r_1(x, x)}{(y - x)^2 - (x - x)^2} \\ &= \frac{D_1 r_1(c_1, x)}{2(c_1 - x)} = \frac{f'(c_1) - f'(x) - f''(x_0)(c_1 - x)}{2(c - x)} \\ &= \frac{r_2(c_1, x)}{2(c - x)} = \frac{r_2(c_1, x) - r_2(x, x)}{2(c - x)} \\ &= \frac{D_1 r_2(c_2, x)}{2} = f''(c_2) - f''(x_0). \end{aligned}$$

Since c_1 is between y and x_0 , and c_2 is between c_1 and x_0 , the continuity of f'' shows that the limit above as $(x, y) \rightarrow (x_0, y_0)$ is 0.

EXERCISE 8.31

We may copy the argument for Exercise 8.8 word by word. The argument using small o notation is equivalent to using Exercises 8.27 and 8.29.

Let

$$\begin{aligned} F(\vec{x}) &= F(\vec{x}_0) + F'(\vec{x}_0)(\Delta\vec{x}) + R_1, & R_1 &= o(\|\Delta\vec{x}\|); \\ G(\vec{x}) &= G(\vec{x}_0) + G'(\vec{x}_0)(\Delta\vec{x}) + R_2, & R_2 &= o(\|\Delta\vec{x}\|). \end{aligned}$$

Then

$$\begin{aligned} B(F(\vec{x}), G(\vec{x})) &= B(F(\vec{x}_0) + F'(\vec{x}_0)(\Delta\vec{x}) + R_1, G(\vec{x}_0) + G'(\vec{x}_0)(\Delta\vec{x}) + R_2) \\ &= B(F(\vec{x}_0), G(\vec{x}_0)) + B(F'(\vec{x}_0)(\Delta\vec{x}), G(\vec{x}_0)) + B(F(\vec{x}_0), G'(\vec{x}_0)(\Delta\vec{x})) + R, \\ R &= B(F(\vec{x}_0), R_2) + B(R_1, G(\vec{x}_0)) + B(F'(\vec{x}_0)(\Delta\vec{x}), R_2) + B(R_1, G'(\vec{x}_0)(\Delta\vec{x})) \\ &\quad + B(F'(\vec{x}_0)(\Delta\vec{x}), G'(\vec{x}_0)(\Delta\vec{x})) + B(R_1, R_2). \end{aligned}$$

By $\|B(\vec{u}, \vec{v})\| \leq \|B\|\|\vec{u}\|\|\vec{v}\|$, we have

$$\begin{aligned} \|B(F(\vec{x}_0), R_2)\| &\leq \|B\|\|F(\vec{x}_0)\|\|R_2\|, \\ \|B(R_1, G(\vec{x}_0))\| &\leq \|B\|\|G(\vec{x}_0)\|\|R_1\|, \\ \|B(F'(\vec{x}_0)(\Delta\vec{x}), R_2)\| &\leq \|B\|\|F'(\vec{x}_0)(\Delta\vec{x})\|\|R_2\| \leq \|B\|\|F'(\vec{x}_0)\|\|\Delta\vec{x}\|\|R_2\|, \\ \|B(R_1, G'(\vec{x}_0)(\Delta\vec{x}))\| &\leq \|B\|\|G'(\vec{x}_0)(\Delta\vec{x})\|\|R_1\| \leq \|B\|\|G'(\vec{x}_0)\|\|\Delta\vec{x}\|\|R_1\|, \\ \|B(F'(\vec{x}_0)(\Delta\vec{x}), G'(\vec{x}_0)(\Delta\vec{x}))\| &\leq \|B\|\|F'(\vec{x}_0)(\Delta\vec{x})\|\|G'(\vec{x}_0)(\Delta\vec{x})\| \leq \|B\|\|F'(\vec{x}_0)\|\|G'(\vec{x}_0)\|\|\Delta\vec{x}\|^2, \\ \|B(R_1, R_2)\| &\leq \|B\|\|R_1\|\|R_2\|. \end{aligned}$$

Then by $R_1 = o(\Delta\vec{x})$ and $R_2 = o(\Delta\vec{x})$, the estimations above imply that $B(F(\vec{x}_0), R_2) = o(\|\Delta\vec{x}\|)$, $B(F'(\vec{x}_0), R_2) = o(\|\Delta\vec{x}\|)$, and the other four terms in R are $o(\|\Delta\vec{x}\|^2)$. Therefore $R = o(\|\Delta\vec{x}\|)$, and $B(F(\vec{x}_0), G(\vec{x}_0)) + B(F'(\vec{x}_0)(\Delta\vec{x}), G(\vec{x}_0)) + B(F(\vec{x}_0), G'(\vec{x}_0)(\Delta\vec{x}))$ is the linear approximation of $B(F(\vec{x}), G(\vec{x}))$ at \vec{x}_0 . In particular, we get

$$B(F, G)'(\vec{x}_0)(\vec{v}) = B(F'(\vec{x}_0)(\vec{v}), G(\vec{x}_0)) + B(F(\vec{x}_0), G'(\vec{x}_0)(\vec{v})).$$

[Alternative: Argue as special case of Exercise 8.28]

Let $\vec{a} = F(\vec{x}_0)$, $\vec{b} = G(\vec{x}_0)$, $L = F'(\vec{x}_0)$ and $K = G'(\vec{x}_0)$. Then

$$F(\vec{x}) \sim_{\|\Delta\vec{x}\|} \vec{a} + L(\Delta\vec{x}), \quad G(\vec{x}) \sim_{\|\Delta\vec{x}\|} \vec{b} + K(\Delta\vec{x}).$$

By Exercise 8.28, we have

$$\begin{aligned} B(F(\vec{x}), G(\vec{x})) &\sim_{\|\Delta\vec{x}\|} B(\vec{a} + L(\Delta\vec{x}), \vec{b} + K(\Delta\vec{x})) \\ &= B(\vec{a}, \vec{b}) + B(L(\Delta\vec{x}), \vec{b}) + B(\vec{a}, K(\Delta\vec{x})) + B(L(\Delta\vec{x}), K(\Delta\vec{x})). \end{aligned}$$

Since

$$\|B(L(\Delta\vec{x}), K(\Delta\vec{x}))\| \leq \|B\|\|L(\Delta\vec{x})\|\|K(\Delta\vec{x})\| \leq \|B\|\|L\|\|K\|\|\Delta\vec{x}\|^2,$$

we get $B(L(\Delta\vec{x}), K(\Delta\vec{x})) \sim_{\|\Delta\vec{x}\|} 0$. Then by Exercise 8.27, we get

$$\begin{aligned} & B(\vec{a}, \vec{b}) + B(L(\Delta\vec{x}), \vec{b}) + B(\vec{a}, K(\Delta\vec{x})) + B(L(\Delta\vec{x}), K(\Delta\vec{x})) \\ & \sim_{\|\Delta\vec{x}\|} B(\vec{a}, \vec{b}) + B(L(\Delta\vec{x}), \vec{b}) + B(\vec{a}, K(\Delta\vec{x})). \end{aligned}$$

Further by Exercise 8.29, we get

$$B(F(\vec{x}), G(\vec{x})) \sim_{\|\Delta\vec{x}\|} B(\vec{a}, \vec{b}) + B(L(\Delta\vec{x}), \vec{b}) + B(\vec{a}, K(\Delta\vec{x})).$$

Then the linear part

$$B(L(\Delta\vec{x}), \vec{b}) + B(\vec{a}, K(\Delta\vec{x})) = B(F'(\vec{x}_0)(\Delta\vec{x}), G(\vec{x}_0)) + B(F(\vec{x}_0), G'(\vec{x}_0)(\Delta\vec{x}))$$

of the right side is then the derivative $B(F, G)'(\vec{x}_0)(\Delta\vec{x})$.

EXERCISE 8.32

If G is a multilinear map, and F_1, \dots, F_k are differentiable maps, then

$$G(F_1, F_2, \dots, F_k)' = G(F_1', F_2, \dots, F_k) + G(F_1, F_2', \dots, F_k) + \dots + G(F_1, F_2, \dots, F_k').$$

EXERCISE 8.33 (1)

We have

$$F(\vec{x}) = F(\vec{x}_0) + F'(\vec{x}_0)(\Delta\vec{x}) + R_1(\vec{x}) = F'(\vec{x}_0)(\Delta\vec{x}) + R_1(\vec{x}), \quad R_1(\vec{x}) = o(\|\Delta\vec{x}\|),$$

and similarly

$$G(\vec{x}) = G(\vec{x}_0) + G'(\vec{x}_0)(\Delta\vec{x}) + R_2(\vec{x}) = G'(\vec{x}_0)(\Delta\vec{x}) + R_2(\vec{x}), \quad R_2(\vec{x}) = o(\|\Delta\vec{x}\|).$$

Then

$$\begin{aligned} B(F(\vec{x}), G(\vec{x})) &= B(F'(\vec{x}_0)(\Delta\vec{x}), G'(\vec{x}_0)(\Delta\vec{x})) \\ &\quad + B(F'(\vec{x}_0)(\Delta\vec{x}), R_2(\vec{x})) + B(R_1(\vec{x}), G'(\vec{x}_0)(\Delta\vec{x})) + B(R_1(\vec{x}), R_2(\vec{x})). \end{aligned}$$

For any $\epsilon > 0$, there is $\delta > 0$, such that

$$\|\Delta\vec{x}\| < \delta \implies \|R_1(\vec{x})\| < \epsilon\|\Delta\vec{x}\|, \quad \|R_2(\vec{x})\| < \epsilon\|\Delta\vec{x}\|.$$

Then

$$\begin{aligned} \|B(F'(\vec{x}_0)(\Delta\vec{x}), R_2(\vec{x}))\| &\leq \|B\| \|F'(\vec{x}_0)(\Delta\vec{x})\| \|R_2(\vec{x})\| \leq \|B\| \|F'(\vec{x}_0)\| \|\Delta\vec{x}\| \epsilon \|\Delta\vec{x}\|, \\ \|B(R_1(\vec{x}), G'(\vec{x}_0)(\Delta\vec{x}))\| &\leq \|B\| \|R_1(\vec{x})\| \|G'(\vec{x}_0)(\Delta\vec{x})\| \leq \|B\| \epsilon \|\Delta\vec{x}\| \|G'(\vec{x}_0)\| \|\Delta\vec{x}\|, \\ \|B(R_1(\vec{x}), R_2(\vec{x}))\| &\leq \|B\| \|R_1(\vec{x})\| \|R_2(\vec{x})\| \leq \|B\| \epsilon \|\Delta\vec{x}\| \epsilon \|\Delta\vec{x}\|. \end{aligned}$$

This implies

$$\|B(F(\vec{x}), G(\vec{x})) - B(F'(\vec{x}_0)(\Delta\vec{x}), G'(\vec{x}_0)(\Delta\vec{x}))\| \leq \epsilon \|B\| (\|F'(\vec{x}_0)\| + \|G'(\vec{x}_0)\| + \epsilon) \|\Delta\vec{x}\|,$$

and proves

$$B(F, G) \sim_{\|\Delta\vec{x}\|^2} B(F'(\vec{x}_0)(\Delta x), G'(\vec{x}_0)(\Delta x)).$$

In general, if $G(\vec{y}_1, \dots, \vec{y}_k)$ is multilinear, and $F_i(\vec{x}_0) = \vec{0}$, then

$$G(F_1, \dots, F_k) \sim_{\|\Delta\vec{x}\|^k} G(F'_1(\vec{x}_0)(\Delta x), \dots, F'_k(\vec{x}_0)(\Delta x)).$$

EXERCISE 8.33 (2)

We have $Q(\vec{y}) = B(\vec{y}, \vec{y})$ for a symmetric bilinear map B . Then by the first part, we have

$$Q(F) \sim_{\|\Delta\vec{x}\|^2} B(F'(\vec{x}_0)(\Delta x), F'(\vec{x}_0)(\Delta x)) = Q(F'(\vec{x}_0)(\Delta x)).$$

In general, if $G(\vec{y})$ is k -th order, and $F_i(\vec{x}_0) = \vec{0}$, then

$$G(F(\vec{x})) \sim_{\|\Delta\vec{x}\|^k} G(F'(\vec{x}_0)(\Delta x)).$$

EXERCISE 8.34

The gradient is defined by $f'(\vec{x}_0)(\vec{v}) = \nabla f(\vec{x}_0) \cdot \vec{v}$ for all \vec{v} . The calculation of the gradient follows from the calculation of the derivative.

By $(f + g)'(\vec{x}_0) = f'(\vec{x}_0) + g'(\vec{x}_0)$, we get

$$\nabla(f + g)(\vec{x}_0) \cdot \vec{v} = \nabla f(\vec{x}_0) \cdot \vec{v} + \nabla g(\vec{x}_0) \cdot \vec{v} = (\nabla f(\vec{x}_0) + \nabla g(\vec{x}_0)) \cdot \vec{v}.$$

Since this holds for all \vec{v} , we get $\nabla(f + g) = \nabla f + \nabla g$.

By $(fg)'(\vec{x}_0)(\vec{v}) = g(\vec{x}_0)f'(\vec{x}_0)(\vec{v}) + f(\vec{x}_0)g'(\vec{x}_0)(\vec{v})$, we get

$$\nabla(fg)(\vec{x}_0) \cdot \vec{v} = g(\vec{x}_0)(\nabla f(\vec{x}_0) \cdot \vec{v}) + f(\vec{x}_0)(\nabla g(\vec{x}_0) \cdot \vec{v}) = (g(\vec{x}_0)\nabla f(\vec{x}_0) + f(\vec{x}_0)\nabla g(\vec{x}_0)) \cdot \vec{v}.$$

Since this holds for all \vec{v} , we get $\nabla(fg) = g\nabla f + f\nabla g$.

We have

$$\begin{aligned} (F \cdot G)'(\vec{x}_0)(\vec{v}) &= F'(\vec{x}_0)(\vec{v}) \cdot G(\vec{x}_0) + F(\vec{x}_0) \cdot G'(\vec{x}_0)(\vec{v}) \\ &= \vec{v} \cdot F'(\vec{x}_0)^*(G(\vec{x}_0)) + G'(\vec{x}_0)^*(F(\vec{x}_0)) \cdot \vec{v} \\ &= [F'(\vec{x}_0)^*(G(\vec{x}_0)) + G'(\vec{x}_0)^*(F(\vec{x}_0))] \cdot \vec{v}. \end{aligned}$$

Since this holds for all \vec{v} , we get

$$\nabla(F \cdot G)(\vec{x}_0) = F'(\vec{x}_0)^*(G(\vec{x}_0)) + G'(\vec{x}_0)^*(F(\vec{x}_0)),$$

or

$$\nabla(F \cdot G) = F'^*(G) + G'^*(F).$$

EXERCISE 8.35

We may prove the chain rule similar to the single variable case. This means applying Exercise 8.30 to the case P is the linear approximation of F , with $u = \|\Delta\vec{x}\|$, and Q is the linear approximation of G , with $v = \|\Delta\vec{y}\|$.

The following is a more direct proof.

Let $\vec{y}_0 = F(\vec{x}_0)$, $\vec{z}_0 = G(\vec{y}_0)$, $L = F'(\vec{x}_0)$, $K = G'(\vec{y}_0)$. For any $\epsilon > 0$, there is $\mu > 0$, such that $\|\Delta\vec{y}\| < \mu$ implies

$$\|G(\vec{y}) - \vec{z}_0 - K(\Delta\vec{y})\| \leq \epsilon\|\Delta\vec{y}\|.$$

Then there is $\delta > 0$, such that $\|\Delta\vec{x}\| < \delta$ implies

$$\|F(\vec{x}) - \vec{y}_0 - L(\Delta\vec{x})\| \leq \epsilon\|\Delta\vec{x}\|.$$

Note that this implies $\|F(\vec{x}) - \vec{y}_0\| \leq (\|L\| + \epsilon)\|\Delta\vec{x}\| < (\|L\| + \epsilon)\delta$. By choosing $\delta < \frac{\mu}{\|L\| + \epsilon}$ to begin with, we also know $\|F(\vec{x}) - \vec{y}_0\| < \mu$, so that

$$\|G(F(\vec{x})) - \vec{z}_0 - K(F(\vec{x}) - \vec{y}_0)\| \leq \epsilon\|F(\vec{x}) - \vec{y}_0\| \leq \epsilon(\|L\| + \epsilon)\|\Delta\vec{x}\|.$$

Then

$$\begin{aligned} & \|G(F(\vec{x})) - \vec{z}_0 - K(L(\Delta\vec{x}))\| \\ & \leq \|G(F(\vec{x})) - \vec{z}_0 - K(F(\vec{x}) - \vec{y}_0)\| + \|K(F(\vec{x}) - \vec{y}_0 - L(\Delta\vec{x}))\| \\ & \leq \epsilon(\|L\| + \epsilon)\|\Delta\vec{x}\| + \|K\|\|F(\vec{x}) - \vec{y}_0 - L(\Delta\vec{x})\| \leq \epsilon(\|L\| + \epsilon + \|K\|)\|\Delta\vec{x}\|. \end{aligned}$$

This implies that $\vec{z}_0 + K(L(\Delta\vec{x}))$ is a linear approximation of $G(F(\vec{x}))$ at \vec{x}_0 and $(G \circ F)'(\vec{x}_0) = K \circ L$.

EXERCISE 8.36

By restricting (chain rule used here) to $\vec{x} = t\vec{v}$, we see that $f(\|\vec{x}\|_2)$ is differentiable away from $\vec{0}$ if and only if $f(t)$ is differentiable for $t > 0$.

Now consider the differentiability of $f(\|\vec{x}\|_2)$ at $\vec{0}$. First, since differentiable functions are continuous, we must have $f(0) = \lim_{t \rightarrow 0^+} f(t)$. Suppose the linear functional $l(\vec{v})$ is the derivative of $f(\|\vec{x}\|_2)$ at $\vec{0}$. Then for any $\epsilon > 0$, there is $\delta > 0$, such that $\|\vec{x}\|_2 < \delta$ implies $|f(\|\vec{x}\|_2) - f(0) - l(\vec{x})| \leq \epsilon\|\vec{x}\|_2$. By applying the condition to $-\vec{x}$, we find $\|\vec{x}\|_2 < \delta$ implies $|f(\|\vec{x}\|_2) - f(0) + l(\vec{x})| \leq \epsilon\|\vec{x}\|_2$. Thus $\|\vec{x}\|_2 < \delta$ implies $|2l(\vec{x})| \leq |f(\|\vec{x}\|_2) - f(0) - l(\vec{x})| + |f(\|\vec{x}\|_2) - f(0) + l(\vec{x})| \leq 2\epsilon\|\vec{x}\|_2$. Thus $|l(\vec{x})| \leq \epsilon\|\vec{x}\|_2$ for any \vec{x} satisfying $\|\vec{x}\|_2 < \delta$. This further implies $|l(c\vec{x})| \leq \epsilon\|c\vec{x}\|_2$, where $c\vec{x}$ can be any vector of arbitrary length. Therefore we must have $l(c\vec{x}) = 0$ for all \vec{x} .

Thus the differentiability at $\vec{0}$ means that for any $\epsilon > 0$, there is $\delta > 0$, such that $\|\vec{x}\|_2 < \delta$ implies $|f(\|\vec{x}\|_2) - f(0)| \leq \epsilon\|\vec{x}\|_2$. In other words, $0 < t < \delta$ implies $|f(t) - f(0)| \leq \epsilon t$. This means $\lim_{t \rightarrow 0^+} \frac{f(t) - f(0)}{t} = 0$. Thus we conclude that $f(\|\vec{x}\|_2)$ is differentiable at $\vec{0}$ if and only if $f'_+(0) = 0$.

EXERCISE 8.37

By the chain rule, $F \circ G = id$ and $G \circ F = id$ implies $F' \circ G' = I$ and $G' \circ F' = I$. Thus the linear transforms F' and G' are inverse to each other.

EXERCISE 8.38

By taking the derivative of $f(t, t^2) = 1$, we get $f_x(t, t^2) + f_y(t, t^2)2t = 0$. Then by $f_x(t, t^2) = t$, we get $f_y(t, t^2) = -\frac{1}{2}$.

EXERCISE 8.39

Let $\nabla f = (a, b)$. Then $\phi' = (1, 2)$ and $\psi' = (2, 1)$ at $(1, 1)$, and the condition tells us

$$2 = \nabla f \cdot (1, 2) = a + 2b, \quad 3 = \nabla f \cdot (2, 1) = 2a + b.$$

Solving the system, we get $\nabla f = \frac{1}{3}(4, 1)$.

EXERCISE 8.40

We have $f_r = f_x \cos \theta + f_y \sin \theta$ and $f_\theta = -f_x r \sin \theta + f_y \cos \theta$. Then $D_{\vec{e}_r} f = \nabla f \cdot \vec{e}_r = f_x \cos \theta + f_y \sin \theta = f_r$ and $D_{\vec{e}_\theta} f = \nabla f \cdot \vec{e}_\theta = -f_x \sin \theta + f_y \cos \theta = r^{-1} f_\theta$. By Exercise 6.1.32, we have $\nabla f = f_r \vec{e}_r + r^{-1} f_\theta \vec{e}_\theta$.

EXERCISE 8.41

We have $f_r = f_x \cos \theta + f_y \sin \theta = r^{-1}(x f_x + y f_y)$. Therefore f is independent of $r \iff f_r = 0 \iff x f_x + y f_y = 0$.

We also have $f_\theta = -f_x r \sin \theta + f_y r \cos \theta = -y f_x + x f_y$. Therefore f is independent of $\theta \iff f_\theta = 0 \iff -y f_x + x f_y = 0$.

EXERCISE 8.42

$$\begin{aligned} x g_x + y g_y + z g_z &= x \left(f_v \frac{\sqrt{z}}{2\sqrt{x}} + f_w \frac{\sqrt{y}}{2\sqrt{x}} \right) + y \left(f_w \frac{\sqrt{x}}{2\sqrt{y}} + f_u \frac{\sqrt{z}}{2\sqrt{y}} \right) + z \left(f_u \frac{\sqrt{y}}{2\sqrt{z}} + f_v \frac{\sqrt{x}}{2\sqrt{z}} \right) \\ &= \left(y \frac{\sqrt{z}}{2\sqrt{y}} + z \frac{\sqrt{y}}{2\sqrt{z}} \right) f_u + \left(z \frac{\sqrt{x}}{2\sqrt{z}} + x \frac{\sqrt{z}}{2\sqrt{x}} \right) f_v + \left(x \frac{\sqrt{y}}{2\sqrt{x}} + y \frac{\sqrt{x}}{2\sqrt{y}} \right) f_w \\ &= \sqrt{yz} f_u + \sqrt{zx} f_v + \sqrt{xy} f_w = u f_u + v f_v + w f_w. \end{aligned}$$

EXERCISE 8.43

If $f(x, y) = \phi(xy)$, then $f_x = y\phi'(xy)$, $f_y = x\phi'(xy)$. Therefore $x f_x = y f_y$.

If $f(x, y) = \phi\left(\frac{y}{x}\right)$, then $f_x = -\frac{y}{x^2}\phi'\left(\frac{y}{x}\right)$, $f_y = \frac{1}{x}\phi'\left(\frac{y}{x}\right)$. Therefore $x f_x + y f_y = 0$.

EXERCISE 8.44

We have

$$(g \circ f)'(\vec{x}_0)(\vec{v}) = g'(f(\vec{x}_0))(f'(\vec{x}_0)(\vec{v})) = g'(f(\vec{x}_0))(\nabla f(\vec{x}_0) \cdot \vec{v}) = [g'(f(\vec{x}_0))\nabla f(\vec{x}_0)] \cdot \vec{v}.$$

Here the last equality is due to the following reason: Both $g'(f(\vec{x}_0))$ and $\nabla f(\vec{x}_0) \cdot \vec{v}$ are actually numbers. Therefore the application of the former (as linear transform) to the later (as vector) is simply multiplication of two numbers. Since the equality holds for all \vec{v} , we get

$$\nabla(g \circ f)(\vec{x}_0) = g'(f(\vec{x}_0))\nabla f(\vec{x}_0), \quad \text{or} \quad \nabla(g \circ f) = (g' \circ f)\nabla f.$$

For the more general $g(F(\vec{x}))$, we have

$$(g \circ F)'(\vec{x}_0)(\vec{v}) = g'(F(\vec{x}_0))(F'(\vec{x}_0)(\vec{v})) = \nabla g(F(\vec{x}_0)) \cdot F'(\vec{x}_0)(\vec{v}) = F'(\vec{x}_0)^*(\nabla g(F(\vec{x}_0))) \cdot \vec{v}.$$

Since the equality holds for all \vec{v} , we get

$$\nabla(g \circ F)(\vec{x}_0) = F'(\vec{x}_0)^*(\nabla g(F(\vec{x}_0))), \text{ or } \nabla(g \circ F) = F'^*(\nabla g \circ F).$$

EXERCISE 8.45

The derivatives of $X \rightarrow A^{-1}X$, $X \rightarrow XA^{-1}$ are $H \rightarrow A^{-1}H$, $H \rightarrow HA^{-1}$ because the two maps are linear. By Exercise 8.12, the derivative of $X \rightarrow X^{-1}$ at I is $H \rightarrow -H$. Thus the composition $X \rightarrow (A^{-1}X)^{-1}A^{-1} = X^{-1}$ has the derivative $(-A^{-1}H)A^{-1} = -A^{-1}HA^{-1}$ at A .

EXERCISE 8.46

The function is $f(x, y) = \frac{x^2y}{x^2 + y^2}$ for $(x, y) \neq (0, 0)$ and $f(0, 0) = 0$. Since $\nabla f(0, 0) = (0, 0)$, we always have $\nabla f(0, 0) \cdot \vec{v} = 0$. On the other hand, the direct calculation shows that $D_{\vec{v}}f = a^2b \neq 0$ is not zero for most $\vec{v} = (a, b)$. So the function fails the formula $D_{\vec{v}}f = \nabla f \cdot \vec{v}$.

Since the formula $D_{\vec{v}}f = \nabla f \cdot \vec{v}$ is a special case of the chain rule, the failure of the formula $D_{\vec{v}}f = \nabla f \cdot \vec{v}$ implies the failure of the chain rule. Since the straight line is always differentiable, the failure must be due to the non-differentiability of f .

EXERCISE 8.47

As explained in Example 8.1.13, the function in Example 6.2.3 satisfies the equality $D_{\vec{v}}f(\vec{x}_0) = \nabla f(\vec{x}_0) \cdot \vec{v}$ for all \vec{v} . As explained in Exercise 8.46, this means that $f(\phi(t))$ satisfies the chain rule for all straight lines $\phi(t)$. Yet the function is not differentiable.

EXERCISE 8.48

The chain rule formula is a formula for the partial derivatives. The partial derivative is taken for a specific variable, while keeping all the other variables constant. Therefore in the chain rule formula for $F \circ G$, only the differentiability of G along one particular variable is used. Since this differentiability is equivalent to the existence of the corresponding partial derivative, we do not expect a counterexample where G only has partial derivatives but is not differentiable.

EXERCISE 8.49

By the equivalence of norms, we will conclude $\|F(\vec{b}) - F(\vec{a})\| \leq C\|F'(\vec{c})\|\|\vec{b} - \vec{a}\|$ for some $C > 0$ in Proposition 8.2.2, where C is independent of the functions and variables (it depends only on the norms of the Euclidean spaces).

EXERCISE 8.50

If $f_x(x, y) = 0$ on an open subset U , such that the intersection of U with any x -line $y = c$ is an interval, then $f(x, y) = g(y)$ for a function g of y .

More generally, suppose $f_{\vec{x}}(\vec{x}, \vec{y})$ on an open subset U , such that the intersection of U with any plane $\vec{y} = \vec{c}$ is path connected, then $f(\vec{x}, \vec{y}) = g(\vec{y})$ for a function g of \vec{y} .