

## Solutions to Review Problems for Exam 1

1. Compute the (shortest) distance from the point  $P(4, 0, -7)$  in  $\mathbb{R}^3$  to the plane given by

$$4x - y - 3z = 12.$$

**Solution:** The point  $P_o(3, 0, 0)$  is in the plane. Let

$$w = \overrightarrow{P_oP} = \begin{pmatrix} 1 \\ 0 \\ -7 \end{pmatrix}$$

The vector  $n = \begin{pmatrix} 4 \\ -1 \\ -3 \end{pmatrix}$  is orthogonal to the plane. To find the shortest distance,  $d$ , from  $P$  to the plane, we compute the norm of the orthogonal projection of  $w$  onto  $n$ ; that is,

$$d = \|\text{Proj}_{\hat{n}}(w)\|,$$

where

$$\hat{n} = \frac{1}{\sqrt{26}} \begin{pmatrix} 4 \\ -1 \\ -3 \end{pmatrix},$$

a unit vector in the direction of  $n$ , and

$$\text{Proj}_{\hat{n}}(w) = (w \cdot \hat{n})\hat{n}.$$

It then follows that

$$d = |w \cdot \hat{n}|,$$

where  $w \cdot \hat{n} = \frac{1}{\sqrt{26}}(4 + 21) = \frac{25}{\sqrt{26}}$ . Hence,  $d = \frac{25\sqrt{26}}{26} \approx 4.9$ .  $\square$

2. Compute the (shortest) distance from the point  $P(4, 0, -7)$  in  $\mathbb{R}^3$  to the line given by the parametric equations

$$\begin{cases} x = -1 + 4t, \\ y = -7t, \\ z = 2 - t. \end{cases}$$

**Solution:** The point  $P_o(-1, 0, 2)$  is on the line. The vector

$$v = \begin{pmatrix} 4 \\ -7 \\ -1 \end{pmatrix}$$

gives the direction of the line. Put

$$w = \overrightarrow{P_oP} = \begin{pmatrix} 5 \\ 0 \\ -9 \end{pmatrix}.$$

The vectors  $v$  and  $w$  determine a parallelogram whose area is the norm of  $v$  times the shortest distance,  $d$ , from  $P$  to the line determined by  $v$  at  $P_o$ . We then have that

$$\text{area}(P(v, w)) = \|v\|d,$$

from which we get that

$$d = \frac{\text{area}(P(v, w))}{\|v\|}.$$

On the other hand,

$$\text{area}(P(v, w)) = \|v \times w\|,$$

where

$$v \times w = \begin{vmatrix} \hat{i} & \hat{j} & \hat{k} \\ 4 & -7 & -1 \\ 5 & 0 & -9 \end{vmatrix} = 63\hat{i} + 31\hat{j} - 35\hat{k}.$$

Thus,  $\|v \times w\| = \sqrt{(63)^2 + (31)^2 + (35)^2} = \sqrt{6155}$  and therefore

$$d = \frac{\sqrt{6155}}{\sqrt{66}} \approx 9.7.$$

□

3. Compute the area of the triangle whose vertices in  $\mathbb{R}^3$  are the points  $(1, 1, 0)$ ,  $(2, 0, 1)$  and  $(0, 3, 1)$

**Solution:** Label the points  $P_o(1, 1, 0)$ ,  $P_1(2, 0, 1)$  and  $P_2(0, 3, 1)$  and define the vectors

$$v = \overrightarrow{P_oP_1} = \begin{pmatrix} 1 \\ -1 \\ 1 \end{pmatrix} \quad \text{and} \quad w = \overrightarrow{P_oP_2} = \begin{pmatrix} -1 \\ 2 \\ 1 \end{pmatrix}.$$

The area of the triangle determined by the points  $P_o$ ,  $P_1$  and  $P_2$  is then half of the area of the parallelogram determined by the vectors  $v$  and  $w$ . Thus,

$$\text{area}(\triangle P_oP_1P_2) = \frac{1}{2}\|v \times w\|,$$

where

$$v \times w = \begin{vmatrix} \hat{i} & \hat{j} & \hat{k} \\ 1 & -1 & 1 \\ -1 & 2 & 1 \end{vmatrix} = -3\hat{i} - 2\hat{j} + \hat{k}.$$

Consequently,  $\text{area}(\triangle P_oP_1P_2) = \frac{1}{2}\sqrt{9 + 4 + 1} = \frac{\sqrt{14}}{2} \approx 1.87$ .  $\square$

4. Let  $v$  and  $w$  be two vectors in  $\mathbb{R}^3$ , and let  $\lambda$  be a scalar. Show that the area of the parallelogram determined by the vectors  $v$  and  $w + \lambda v$  is the same as that determined by  $v$  and  $w$ .

**Solution:** The area of the parallelogram determined by  $v$  and  $w + \lambda v$  is

$$\text{area}(P(v, w + \lambda v)) = \|v \times (w + \lambda v)\|,$$

where

$$v \times (w + \lambda v) = v \times w + \lambda v \times v = v \times w.$$

Consequently,  $\text{area}(P(v, w + \lambda v)) = \|v \times w\| = \text{area}(P(v, w))$ .  $\square$

5. Let  $\hat{u}$  denote a unit vector in  $\mathbb{R}^n$  and  $P_{\hat{u}}(v)$  denote the orthogonal projection of  $v$  along the direction of  $\hat{u}$  for any vector  $v \in \mathbb{R}^n$ . Use the Cauchy-Schwarz inequality to prove that the map

$$v \mapsto P_{\hat{u}}(v) \quad \text{for all } v \in \mathbb{R}^n$$

is a continuous map from  $\mathbb{R}^n$  to  $\mathbb{R}^n$ .

**Solution:**  $P_{\hat{u}}(v) = (v \cdot \hat{u})\hat{u}$  for all  $v \in \mathbb{R}^n$ . Consequently, for any  $w, v \in \mathbb{R}^n$ ,

$$\begin{aligned} P_{\hat{u}}(w) - P_{\hat{u}}(v) &= (w \cdot \hat{u})\hat{u} - (v \cdot \hat{u})\hat{u} \\ &= (w \cdot \hat{u} - v \cdot \hat{u})\hat{u} \\ &= [(w - v) \cdot \hat{u}]\hat{u}. \end{aligned}$$

It then follows that

$$\|P_{\hat{u}}(w) - P_{\hat{u}}(v)\| = |(w - v) \cdot \hat{u}|,$$

since  $\|\hat{u}\| = 1$ . Hence, by the Cauchy–Schwarz inequality,

$$\|P_{\hat{u}}(w) - P_{\hat{u}}(v)\| \leq \|w - v\|.$$

Applying the Squeeze Theorem we then get that

$$\lim_{\|w-v\| \rightarrow 0} \|P_{\hat{u}}(w) - P_{\hat{u}}(v)\| = 0,$$

which shows that  $P_{\hat{u}}$  is continuous at every  $v \in V$ . □

6. Define the scalar field  $f: \mathbb{R}^n \rightarrow \mathbb{R}$  by  $f(v) = \frac{1}{2}\|v\|^2$  for all  $v \in \mathbb{R}^n$ . Show that  $f$  is differentiable on  $\mathbb{R}^n$  and compute the linear map  $Df(u): \mathbb{R}^n \rightarrow \mathbb{R}$  for all  $u \in \mathbb{R}^n$ . What is the gradient of  $f$  at  $u$  for all  $x \in \mathbb{R}^n$ ?

**Solution:** Let  $u$  and  $w$  be any vector in  $\mathbb{R}^n$  and consider

$$\begin{aligned} f(u + w) &= \frac{1}{2}\|u + w\|^2 \\ &= \frac{1}{2}(u + w) \cdot (u + w) \\ &= \frac{1}{2}u \cdot u + u \cdot w + \frac{1}{2}w \cdot w \\ &= \frac{1}{2}\|u\|^2 + u \cdot w + \frac{1}{2}\|w\|^2. \end{aligned}$$

Thus,

$$f(u + w) - f(u) - u \cdot w = \frac{1}{2}\|w\|^2.$$

Consequently,

$$\frac{|f(u+w) - f(u) - u \cdot w|}{\|w\|} = \frac{1}{2}\|w\|,$$

from which we get that

$$\lim_{\|w\| \rightarrow 0} \frac{|f(u+w) - f(u) - u \cdot w|}{\|w\|} = 0,$$

and therefore  $f$  is differentiable at  $u$  with derivative map  $Df(u)$  given by

$$Df(u)w = u \cdot w \quad \text{for all } w \in \mathbb{R}^n.$$

Hence,  $\nabla f(u) = u$  for all  $u \in \mathbb{R}^n$ . □

7. Let  $g: [0, \infty) \rightarrow \mathbb{R}$  be a differentiable, real-valued function of a single variable, and let  $f(x, y) = g(r)$  where  $r = \sqrt{x^2 + y^2}$ .

- (a) Compute  $\frac{\partial r}{\partial x}$  in terms of  $x$  and  $r$ , and  $\frac{\partial r}{\partial y}$  in terms of  $y$  and  $r$ .

**Solution:** Take the partial derivative of  $r^2 = x^2 + y^2$  on both sides with respect to  $x$  to obtain

$$\frac{\partial(r^2)}{\partial x} = 2x.$$

Applying the chain rule on the left-hand side we get

$$2r \frac{\partial r}{\partial x} = 2x,$$

which leads to

$$\frac{\partial r}{\partial x} = \frac{x}{r}.$$

Similarly,  $\frac{\partial r}{\partial y} = \frac{y}{r}$ . □

- (b) Compute  $\nabla f$  in terms of  $g'(r)$ ,  $r$  and the vector  $\mathbf{r} = x\hat{i} + y\hat{j}$ .

**Solution:** Take the partial derivative of  $f(x, y) = g(r)$  on both sides with respect to  $x$  and apply the Chain Rule to obtain

$$\frac{\partial f}{\partial x} = g'(r) \frac{\partial r}{\partial x} = g'(r) \frac{x}{r}.$$

Similarly,  $\frac{\partial f}{\partial y} = g'(r)\frac{y}{r}$ .

It then follows that

$$\begin{aligned}\nabla f &= \frac{\partial f}{\partial x}\hat{i} + \frac{\partial f}{\partial y}\hat{j} \\ &= g'(r)\frac{x}{r}\hat{i} + g'(r)\frac{y}{r}\hat{j} \\ &= \frac{g'(r)}{r}(x\hat{i} + y\hat{j}) \\ &= \frac{g'(r)}{r}\mathbf{r}.\end{aligned}$$

□

8. Let  $f: U \rightarrow \mathbb{R}$  denote a scalar field defined on an open subset  $U$  of  $\mathbb{R}^n$ , and let  $\hat{u}$  be a unit vector in  $\mathbb{R}^n$ . If the limit

$$\lim_{t \rightarrow 0} \frac{f(v + t\hat{u}) - f(v)}{t}$$

exists, we call it the *directional derivative of  $f$  at  $v$  in the direction of the unit vector  $\hat{u}$* . We denote it by  $D_{\hat{u}}f(v)$ .

- (a) Show that if  $f$  is differentiable at  $v \in U$ , then, for any unit vector  $\hat{u}$  in  $\mathbb{R}^n$ , the directional derivative of  $f$  in the direction of  $\hat{u}$  at  $v$  exists, and

$$D_{\hat{u}}f(v) = \nabla f(v) \cdot \hat{u},$$

where  $\nabla f(v)$  is the gradient of  $f$  at  $v$ .

*Proof:* Suppose that  $f$  is differentiable at  $v \in U$ . Then,

$$f(v + w) = f(v) + Df(v)w + E(w),$$

where

$$Df(v)w = \nabla f(v) \cdot w,$$

and

$$\lim_{\|w\| \rightarrow 0} \frac{|E(w)|}{\|w\|} = 0.$$

Thus, for any  $t \in \mathbb{R}$ ,

$$f(v + t\hat{u}) = f(v) + t\nabla f(v) \cdot \hat{u} + E(t\hat{u}),$$

where

$$\lim_{|t| \rightarrow 0} \frac{|E(t\hat{u})|}{|t|} = 0,$$

since  $\|t\hat{u}\| = |t|\|\hat{u}\| = |t|$ .

We then have that, for  $t \neq 0$ ,

$$\frac{f(v + t\hat{u}) - f(v)}{t} - \nabla f(v) \cdot \hat{u} = \frac{E(t\hat{u})}{t},$$

and consequently

$$\left| \frac{f(v + t\hat{u}) - f(v)}{t} - \nabla f(v) \cdot \hat{u} \right| = \frac{|E(t\hat{u})|}{|t|},$$

from which we get that

$$\lim_{t \rightarrow 0} \left| \frac{f(v + t\hat{u}) - f(v)}{t} - \nabla f(v) \cdot \hat{u} \right| = 0.$$

□

- (b) Suppose that  $f: U \rightarrow \mathbb{R}$  is differentiable at  $v \in U$ . Prove that if  $D_{\hat{u}}f(v) = 0$  for every unit vector  $\hat{u}$  in  $\mathbb{R}^n$ , then  $\nabla f(v)$  must be the zero vector.

*Proof:* Suppose, by way of contradiction, that  $\nabla f(v) \neq \mathbf{0}$ , and put

$$\hat{u} = \frac{1}{\|\nabla f(v)\|} \nabla f(v).$$

Then,  $\hat{u}$  is a unit vector, and therefore, by the assumption,

$$D_{\hat{u}}f(v) = 0,$$

or

$$\nabla f(v) \cdot \hat{u} = 0.$$

But this implies that

$$\nabla f(v) \cdot \frac{1}{\|\nabla f(v)\|} \nabla f(v) = 0,$$

where

$$\begin{aligned} \nabla f(v) \cdot \frac{1}{\|\nabla f(v)\|} \nabla f(v) &= \frac{1}{\|\nabla f(v)\|} \nabla f(v) \cdot \nabla f(v) \\ &= \frac{1}{\|\nabla f(v)\|} \|\nabla f(v)\|^2 \\ &= \|\nabla f(v)\|. \end{aligned}$$

It then follows that  $\|\nabla f(v)\| = 0$ , which contradicts the assumption that  $\nabla f(v) \neq \mathbf{0}$ . Therefore,  $\nabla f(v)$  must be the zero vector.  $\square$

- (c) Suppose that  $f: U \rightarrow \mathbb{R}$  is differentiable at  $v \in U$ . Use the Cauchy–Schwarz inequality to show that the largest value of  $D_{\hat{u}}f(v)$  is  $\|\nabla f(v)\|$  and it occurs when  $\hat{u}$  is in the direction of  $\nabla f(v)$ .

*Proof.* If  $f$  is differentiable at  $x$ , then  $D_{\hat{u}}f(x) = \nabla f(x) \cdot \hat{u}$ , as was shown in part (a). Thus, by the Cauchy–Schwarz inequality,

$$|D_{\hat{u}}f(x)| \leq \|\nabla f(x)\| \|\hat{u}\| = \|\nabla f(x)\|,$$

since  $\hat{u}$  is a unit vector. Hence,

$$-\|\nabla f(x)\| \leq D_{\hat{u}}f(x) \leq \|\nabla f(x)\|$$

for any unit vector  $\hat{u}$ , and so the largest value that  $D_{\hat{u}}f(x)$  can have is  $\|\nabla f(x)\|$ .

If  $\nabla f(x) \neq \mathbf{0}$ , then  $\hat{u} = \frac{1}{\|\nabla f(x)\|} \nabla f(x)$  is a unit vector, and

$$\begin{aligned} D_{\hat{u}}f(x) &= \nabla f(x) \cdot \hat{u} \\ &= \nabla f(x) \cdot \frac{1}{\|\nabla f(x)\|} \nabla f(x) \\ &= \frac{1}{\|\nabla f(x)\|} \nabla f(x) \cdot \nabla f(x) \\ &= \frac{1}{\|\nabla f(x)\|} \|\nabla f(x)\|^2 \\ &= \|\nabla f(x)\|. \end{aligned}$$

Thus,  $D_{\hat{u}}f(x)$  attains its largest value when  $\hat{u}$  is in the direction of  $\nabla f(x)$ .  $\square$



9. The scalar field  $f: U \rightarrow \mathbb{R}$  is said to have a *local minimum* at  $x \in U$  if there exists  $r > 0$  such that  $B_r(x) \subseteq U$  and

$$f(x) \leq f(y) \quad \text{for every } y \in B_r(x).$$

Prove that if  $f$  is differentiable at  $x \in U$  and  $f$  has a local minimum at  $x$ , then  $\nabla f(x) = \mathbf{0}$ , the zero vector in  $\mathbb{R}^n$ .

*Proof.* Let  $\hat{u}$  be a unit vector and  $t \in \mathbb{R}$  be such that  $|t| < r$ ; then,

$$f(x + t\hat{u}) \geq f(x),$$

from which we get that

$$f(x + t\hat{u}) - f(x) \geq 0.$$

Dividing by  $t > 0$  we then have that

$$\frac{f(x + t\hat{u}) - f(x)}{t} \geq 0.$$

Thus, letting  $t \rightarrow 0^+$ , we get that

$$D_{\hat{u}}f(x) \geq 0, \tag{1}$$

since  $f$  is differentiable at  $x$ . Similarly, dividing by  $t < 0$ , we have

$$\frac{f(x + t\hat{u}) - f(x)}{t} \leq 0,$$

from which we obtain, letting  $t \rightarrow 0^-$ , that

$$D_{\hat{u}}f(x) \leq 0. \tag{2}$$

Combining (1) and (2) we then have that

$$D_{\hat{u}}f(x) = 0,$$

where  $\hat{u}$  is an arbitrary unit vector. It then follows from the previous problem that  $\nabla f(x) = \mathbf{0}$ .  $\square$

10. Let  $I$  denote an open interval in  $\mathbb{R}$ . Suppose that  $\sigma: I \rightarrow \mathbb{R}^n$  and  $\gamma: I \rightarrow \mathbb{R}^n$  are paths in  $\mathbb{R}^n$ . Define a real valued function  $f: I \rightarrow \mathbb{R}$  of a single variable by

$$f(t) = \sigma(t) \cdot \gamma(t) \quad \text{for all } t \in I;$$

that is,  $f(t)$  is the dot product of the two paths at  $t$ .

Show that if  $\sigma$  and  $\gamma$  are both differentiable on  $I$ , then so is  $f$ , and

$$f'(t) = \sigma'(t) \cdot \gamma(t) + \sigma(t) \cdot \gamma'(t) \quad \text{for all } t \in I.$$

**Solution:** Let  $t \in I$  and assume that both  $\sigma$  and  $\gamma$  are differentiable at  $t$ . Then,

$$\sigma(t+h) = \sigma(t) + h\sigma'(t) + E_1(h), \quad \text{for } |h| \text{ sufficiently small,}$$

where

$$\lim_{h \rightarrow 0} \frac{\|E_1(h)\|}{|h|} = 0. \quad (3)$$

Similarly,

$$\gamma(t+h) = \gamma(t) + h\gamma'(t) + E_2(h), \quad \text{for } |h| \text{ sufficiently small,}$$

where

$$\lim_{h \rightarrow 0} \frac{\|E_2(h)\|}{|h|} = 0. \quad (4)$$

It then follows that, for  $|h|$  sufficiently small,

$$\begin{aligned} f(t+h) &= \sigma(t+h) \cdot \gamma(t+h) \\ &= (\sigma(t) + h\sigma'(t) + E_1(h)) \cdot (\gamma(t) + h\gamma'(t) + E_2(h)) \\ &= \sigma(t) \cdot \gamma(t) + h\sigma(t) \cdot \gamma'(t) + \sigma(t) \cdot E_2(h) + h\sigma'(t) \cdot \gamma(t) \\ &\quad + h^2\sigma'(t) \cdot \gamma'(t) + h\sigma'(t) \cdot E_2(h) + E_1(h) \cdot \gamma(t) \\ &\quad + hE_1(h) \cdot \gamma'(t) + E_1(h) \cdot E_2(h) \\ &= f(t) + h[\sigma(t) \cdot \gamma'(t) + \sigma'(t) \cdot \gamma(t)] + h^2\sigma'(t) \cdot \gamma'(t) \\ &\quad + \sigma(t) \cdot E_2(h) + h\sigma'(t) \cdot E_2(h) + E_1(h) \cdot \gamma(t) \\ &\quad + hE_1(h) \cdot \gamma'(t) + E_1(h) \cdot E_2(h) \end{aligned}$$

Rearranging terms and dividing by  $h \neq 0$  and  $|h|$  small enough, we then have that

$$\begin{aligned} \frac{f(t+h) - f(t)}{h} &= \sigma(t) \cdot \gamma'(t) + \sigma'(t) \cdot \gamma(t) + h\sigma'(t) \cdot \gamma'(t) \\ &\quad + \sigma(t) \cdot \frac{E_2(h)}{h} + \sigma'(t) \cdot E_2(h) + \frac{E_1(h)}{h} \cdot \gamma(t) \\ &\quad + E_1(h) \cdot \gamma'(t) + E_1(h) \cdot \frac{E_2(h)}{h} \end{aligned}$$

Observe that, as  $h \rightarrow 0$ , all the terms on the right hand side of the previous expression which involve  $E_1$  or  $E_2$  go to 0, by virtue of the

Cauchy–Schwarz inequality and (3) and (4). Therefore, we obtain that

$$\lim_{h \rightarrow 0} \frac{f(t+h) - f(t)}{h} = \sigma(t) \cdot \gamma'(t) + \sigma'(t) \cdot \gamma(t).$$

Hence,  $f$  is differentiable at  $t$ , and its derivative at  $t$  is

$$f'(t) = \sigma(t) \cdot \gamma'(t) + \sigma'(t) \cdot \gamma(t).$$

Since  $t$  is an arbitrary element of  $I$ , the result follows.  $\square$

11. Let  $\sigma: I \rightarrow \mathbb{R}^n$  denote a differentiable path in  $\mathbb{R}^n$ . Show that if  $\|\sigma(t)\|$  is constant for all  $t \in I$ , then  $\sigma'(t)$  is orthogonal to  $\sigma(t)$  for all  $t \in I$ .

**Solution:** Let  $\|\sigma(t)\| = c$ , where  $c$  denotes a constant. Then,

$$\|\sigma(t)\|^2 = c^2,$$

or

$$\sigma(t) \cdot \sigma(t) = c^2.$$

Differentiating with respect to  $t$  on both sides, and using the result of the previous problem, we obtain that

$$\sigma(t) \cdot \sigma'(t) + \sigma'(t) \cdot \sigma(t) = 0,$$

or, by the symmetry of the dot-product,

$$2\sigma'(t) \cdot \sigma(t) = 0,$$

or

$$\sigma'(t) \cdot \sigma(t) = 0.$$

Hence,  $\sigma'(t)$  is orthogonal to  $\sigma(t)$  for all  $t \in I$ .  $\square$

12. A particle is following a path in three-dimensional space given by

$$\sigma(t) = (e^t, e^{-t}, 1 - t) \quad \text{for } t \in \mathbb{R}.$$

At time  $t_0 = 1$ , the particle flies off on a tangent.

- (a) Where will the particle be at time  $t_1 = 2$ ?

**Solution:** Find the tangent line to the path at  $\sigma(1)$ :

$$\vec{r}(t) = \sigma(1) + (t - 1)\sigma'(1),$$

where

$$\sigma'(t) = (e^t, -e^{-t}, -1) \quad \text{for } t \in \mathbb{R}.$$

Then,

$$\vec{r}(t) = (e, 1/e, 0) + (t - 1)(e, -1/e, -1).$$

The parametric equations of the tangent line then are

$$\begin{cases} x = e + e(t - 1) \\ y = 1/e - (t - 1)/e \\ z = 1 - t \end{cases}$$

When  $t = 2$ , the particle will be at the point in  $\mathbb{R}^3$  with coordinates

$$(2e, 0, -1).$$

□

- (b) Will the particle ever hit the  $xy$ -plane? Is so, find the location on the  $xy$  plane where the particle hits.

**Answer:** The particle leaves the path at the point with coordinates  $(e, 1/e, 0)$  on the  $xy$ -plane. After that, it doesn't come back to it. □

13. Let  $U$  denote an open and convex subset of  $\mathbb{R}^n$ . Suppose that  $f: U \rightarrow \mathbb{R}$  is differentiable at every  $x \in U$ . Fix  $x$  and  $y$  in  $U$ , and define  $g: [0, 1] \rightarrow \mathbb{R}$  by

$$g(t) = f(x + t(y - x)) \quad \text{for } 0 \leq t \leq 1.$$

- (a) Explain why the function  $g$  is well defined.

**Solution:** Since  $U$  is convex,  $x + t(y - x)$  is in  $U$  for  $0 \leq t \leq 1$ . Thus,  $f(x + t(y - x))$  is defined for  $t \in [0, 1]$ . □

- (b) Show that  $g$  is differentiable on  $(0, 1)$  and that

$$g'(t) = \nabla f(x + t(y - x)) \cdot (y - x) \quad \text{for } 0 < t < 1.$$

(Suggestion: Consider

$$\frac{g(t+h) - g(t)}{h} = \frac{f(x + t(y-x) + h(y-x)) - f(x + t(y-x))}{h}$$

and apply the definition of differentiability of  $f$  at the point  $x + t(y - x)$ .)

*Proof.* Since  $f$  is differentiable on  $U$ , for  $|h|$  small enough,

$$f(x+t(y-x)+h(y-x)) = f(x+t(y-x)) + Df(x+t(y-x))(h(y-x)) + E(h(y-x)),$$

where

$$\lim_{\|w\| \rightarrow 0} \frac{|E(w)|}{\|w\|} = 0. \quad (5)$$

Thus,

$$f(x+t(y-x)+h(y-x)) = f(x+t(y-x)) + h\nabla f(x+t(y-x)) \cdot (y-x) + E(h(y-x)),$$

from which we get that

$$\begin{aligned} \frac{g(t+h) - g(t)}{h} &= \frac{f(x+t(y-x)+h(y-x)) - f(x+t(y-x))}{h} \\ &= \nabla f(x+t(y-x)) \cdot (y-x) + \frac{E(h(y-x))}{h} \end{aligned}$$

for  $h \neq 0$ .

Observe that

$$\lim_{h \rightarrow 0} \frac{|E(h(y-x))|}{h} = \lim_{h \rightarrow 0} \|y-x\| \frac{|E(h(y-x))|}{\|h(y-x)\|} = 0,$$

by virtue of (5). It then follows that

$$\lim_{h \rightarrow 0} \frac{g(t+h) - g(t)}{h} = \nabla f(x+t(y-x)) \cdot (y-x),$$

and therefore  $g$  is differentiable at  $t$  and  $g'(t) = \nabla f(x+t(y-x)) \cdot (y-x)$ .  $\square$

- (c) Use the Mean Value Theorem for derivatives to show that there exists a point  $z$  on the line segment connecting  $x$  to  $y$  such that

$$f(y) - f(x) = D_{\hat{u}}f(z)\|y-x\|,$$

where  $\hat{u}$  is the unit vector in the direction of the vector  $y-x$ ; that is,  $\hat{u} = \frac{1}{\|y-x\|}(y-x)$ .

(*Hint:* Observe that  $g(1) - g(0) = f(y) - f(x)$ .)

**Solution:** Assume that  $x \neq y$ , for if  $x = y$  the equality certainly holds true.

By the Mean Value Theorem, there exists  $\tau \in (0, 1)$  such that

$$g(1) - g(0) = g'(\tau)(1 - 0) = g'(\tau).$$

It then follows that

$$f(y) - f(x) = \nabla f(x + \tau(y - x)) \cdot (y - x).$$

Put  $z = x + \tau(y - x)$ ; then,  $z$  is a point in the line segment connecting  $x$  to  $y$ , and

$$\begin{aligned} f(y) - f(x) &= \nabla f(z) \cdot (y - x) \\ &= \nabla f(z) \cdot \frac{y - x}{\|y - x\|} \|y - x\| \\ &= \nabla f(z) \cdot \hat{u} \|y - x\| \\ &= D_{\hat{u}}f(z) \|y - x\|, \end{aligned}$$

$$\text{where } \hat{u} = \frac{1}{\|y - x\|} (y - x). \quad \square$$

14. Prove that if  $U$  is an open and convex subset of  $\mathbb{R}^n$ , and  $f: U \rightarrow \mathbb{R}$  is differentiable on  $U$  with  $\nabla f(v) = \mathbf{0}$  for all  $v \in U$ , then  $f$  must be a constant function.

**Solution:** Fix  $x_o \in U$ ; then, since  $U$  is convex, for any  $x \in U \setminus \{x_o\}$ , the line segment connecting  $x_o$  to  $x$  is entirely contained in  $U$ . Furthermore, by the argument in part (c) of the previous problem, there exists  $z$  in the line segment connecting  $x_o$  to  $x$  such that

$$f(x) - f(x_o) = D_{\hat{u}}f(z) \|x - x_o\|,$$

$$\text{where } \hat{u} = \frac{1}{\|x - x_o\|} (x - x_o).$$

Now,  $D_{\hat{u}}f(z) = \nabla f(z) \cdot \hat{u} = 0$ , since  $\nabla f(x) = \mathbf{0}$  for all  $x \in U$ . Therefore,

$$f(x) = f(x_o).$$

Since  $x$  was arbitrary, it follows that  $f$  maps every element in  $U$  to  $f(x_o)$ ; that is,  $f$  is a constant function.  $\square$