

## Solutions to Review Problems for Exam 2

1. 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  $u \in \mathbb{R}^n$ . □

2. 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

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

□

3. 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  $\nabla f(x)$ .  $\square$

4. 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

$$D_{\widehat{u}}f(x) \geq 0, \quad (1)$$

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

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

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

$$D_{\widehat{u}}f(x) \leq 0. \quad (2)$$

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

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

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

5. 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.  $\square$

6. 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  $t \in I$ . □

7. 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.  $\square$

8. 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  $t \in [0, 1]$ .  $\square$

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

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

**Solution:** Apply the Chain Rule to the maps  $f: U \rightarrow \mathbb{R}$  and  $\sigma: (0, 1) \rightarrow \mathbb{R}^n$  given by

$$\sigma(t) = x + t(y - x) \quad \text{for } t \in (0, 1).$$

Since  $U$  is convex, it follows that  $\sigma(t) \in U$  for all  $t \in (0, 1)$ . Consequently,  $\sigma((0, 1)) \subseteq U$  and therefore  $f \circ \sigma: (0, 1) \rightarrow \mathbb{R}$  is defined. Furthermore, by the Chain Rule,  $f \circ \sigma$  is differentiable with

$$D(f \circ \sigma)(t) = Df(\sigma(t))D\sigma(t) = \nabla f(\sigma(t)) \cdot \sigma'(t) \quad \text{for all } t \in (0, 1).$$

Note that  $g = f \circ \sigma$  and  $\sigma'(t) = y - x$  for all  $t \in (0, 1)$ . Hence,

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

which was to be shown.  $\square$

- (c) Use the Mean Value Theorem for derivatives to show that there exists a point  $z$  in 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

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

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

9. 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

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

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. □

10. Let  $f$  be a scalar field defined on  $(x, y)$  where  $x = r \cos \theta$ ,  $y = r \sin \theta$ . Show that

$$\nabla f = \frac{\partial f}{\partial r} \vec{\mathbf{u}}_r + \frac{1}{r} \frac{\partial f}{\partial \theta} \vec{\mathbf{u}}_\theta,$$

where  $\vec{\mathbf{u}}_r = (\cos \theta, \sin \theta)$  and  $\vec{\mathbf{u}}_\theta = (-\sin \theta, \cos \theta)$ .

*Hint:* First find  $\partial f/\partial r$  and  $\partial f/\partial \theta$  in terms of  $\partial f/\partial x$  and  $\partial f/\partial y$  and then solve for  $\partial f/\partial x$  and  $\partial f/\partial y$  in terms of  $\partial f/\partial r$  and  $\partial f/\partial \theta$ .

**Solution:** Given  $f(x, y)$  where  $x = r \cos \theta$  and  $y = r \sin \theta$ , the Chain Rule implies that

$$\frac{\partial f}{\partial r} = \frac{\partial f}{\partial x} \frac{\partial x}{\partial r} + \frac{\partial f}{\partial y} \frac{\partial y}{\partial r}$$

and

$$\frac{\partial f}{\partial \theta} = \frac{\partial f}{\partial x} \frac{\partial x}{\partial \theta} + \frac{\partial f}{\partial y} \frac{\partial y}{\partial \theta},$$

where

$$\frac{\partial x}{\partial r} = \cos \theta,$$

$$\frac{\partial y}{\partial r} = \sin \theta,$$

$$\frac{\partial x}{\partial \theta} = -r \sin \theta,$$

$$\frac{\partial y}{\partial \theta} = r \cos \theta.$$

It then follows that

$$\begin{pmatrix} \frac{\partial f}{\partial r} \\ \frac{\partial f}{\partial \theta} \end{pmatrix} = \begin{pmatrix} \cos \theta \frac{\partial f}{\partial x} + \sin \theta \frac{\partial f}{\partial y} \\ -r \sin \theta \frac{\partial f}{\partial x} + r \cos \theta \frac{\partial f}{\partial y} \end{pmatrix}$$

or

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

Observe that the  $2 \times 2$  matrix  $\begin{pmatrix} \cos \theta & \sin \theta \\ -r \sin \theta & r \cos \theta \end{pmatrix}$  is invertible with inverse

$$\begin{pmatrix} \cos \theta & \sin \theta \\ -r \sin \theta & r \cos \theta \end{pmatrix}^{-1} = \frac{1}{r} \begin{pmatrix} r \cos \theta & -\sin \theta \\ r \sin \theta & \cos \theta \end{pmatrix} = \begin{pmatrix} \cos \theta & -\frac{\sin \theta}{r} \\ \sin \theta & \frac{\cos \theta}{r} \end{pmatrix}.$$

We then have that

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

Which can be written as

$$\begin{pmatrix} \frac{\partial f}{\partial x} \\ \frac{\partial f}{\partial y} \end{pmatrix} = \frac{\partial f}{\partial r} \begin{pmatrix} \cos \theta \\ \sin \theta \end{pmatrix} + \frac{1}{r} \frac{\partial f}{\partial \theta} \begin{pmatrix} -\sin \theta \\ \cos \theta \end{pmatrix}.$$

Transposing the matrices on both sides yields the result.  $\square$

11. Let  $U$  be an open subset of  $\mathbb{R}^n$  and  $I$  be an open interval. Suppose that  $f: U \rightarrow \mathbb{R}$  is a differentiable scalar field and  $\sigma: I \rightarrow \mathbb{R}^n$  be a differentiable path whose image lies in  $U$ . Suppose also that  $\sigma'(t)$  is never the zero vector. Show that if  $f$  has a local maximum or a local minimum at some point on the path, then  $\nabla f$  is perpendicular to the path at that point.

*Suggestion:* Consider the real valued function of a single variable  $g(t) = f(\sigma(t))$  for all  $t \in I$ .

**Solution:** If  $f$  has a local maximum or minimum at  $\sigma(t_0)$ , then  $g'(t_0) = 0$ , where, by the Chain rule,

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

It then follows that

$$\nabla f(\sigma(t_o)) \cdot \sigma'(t_o) = 0,$$

and, consequently,  $\nabla f(\sigma(t_o))$  is perpendicular to the tangent to the path at  $\sigma(t_o)$ .  $\square$

12. Let  $\sigma: [a, b] \rightarrow \mathbb{R}^n$  be a differentiable, one-to-one path. Suppose also that  $\sigma'(t)$ , is never the zero vector. Let  $h: [c, d] \rightarrow [a, b]$  be a one-to-one and onto map such that  $h'(t) \neq 0$  for all  $t \in [c, d]$ . Define

$$\gamma(t) = \sigma(h(t)) \quad \text{for all } t \in [c, d].$$

$\gamma: [c, d] \rightarrow \mathbb{R}^n$  is called a *reparametrization* of  $\sigma$

- (a) Show that  $\gamma$  is a differentiable, one-to-one path.

**Solution:** Since  $\gamma = \sigma \circ h$  is the composition of two differentiable maps, it follows from the Chain Rule that  $\gamma$  is differentiable.

To show that  $\gamma$  is one-to-one, suppose that  $\gamma(t_1) = \gamma(t_2)$  for  $t_1$  and  $t_2$  in  $I$ . It then follows that

$$\sigma(h(t_1)) = \sigma(h(t_2)).$$

Thus, since  $\sigma$  is one-to-one,

$$h(t_1) = h(t_2),$$

from which we get that  $t_1 = t_2$  since  $h$  is one-to-one. Consequently,  $\gamma$  is one-to-one.  $\square$

- (b) Compute  $\gamma'(t)$  and show that it is never the zero vector.

**Solution:** By the Chain Rule,

$$\gamma'(t) = h'(t)\sigma'(h(t)) \quad \text{for all } t \in T.$$

Thus, since neither  $h'(t)$  nor  $\sigma'(h(t))$  are zero,  $\gamma'(t)$  is never the zero vector.  $\square$

- (c) Show that  $\sigma$  and  $\gamma$  have the same image in  $\mathbb{R}^n$ .

*Solution:* We show that

$$\sigma([a, b]) = \gamma([c, d]). \quad (5)$$

Let  $x \in \gamma([c, d])$ ; then, there exists  $t \in [c, d]$  such that

$$x = \gamma(t) = \sigma(h(t)).$$

Thus, there exists  $h(t) \in [a, b]$  such that  $x = \sigma(h(t))$ ; that is,  $x \in \sigma([a, b])$ . Hence,

$$\gamma([c, d]) \subseteq \sigma([a, b]). \quad (6)$$

To show the reverse inclusion, let  $x \in \sigma([a, b])$ . Then, there exists  $\tau \in [a, b]$  such that

$$x = \sigma(\tau).$$

Since  $h: [c, d] \rightarrow [a, b]$  is onto, there exists  $t \in [c, d]$  such that  $\tau = h(t)$ . Thus,

$$x = \sigma(h(t)) = \gamma(t),$$

which shows that  $x \in \gamma([c, d])$ . It then, follows that

$$\sigma([a, b]) \subseteq \gamma([c, d]). \quad (7)$$

Combining the inclusions in (6) and (7) we obtain the set equality in (5).  $\square$