

# **MAT257 Assignment 3**

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October 8, 2021

1. If  $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$  is differentiable at some  $a \in \mathbb{R}^n$ , we will prove that  $f$  is also continuous at  $a$ . First, by definition, we have  $f(a+h) - f(a) - Df(a)(h) \in o(h)$ , so:

$$\lim_{h \rightarrow 0} \frac{|f(a+h) - f(a) - Df(a)(h)|}{|h|} = 0. \quad (*)$$

Now, to prove that  $f$  is continuous at  $a$ , suppose we are given any  $\epsilon > 0$ . Then, by  $(*)$ , there exists some  $\delta' > 0$  such that all  $h \in \mathbb{R}^n$  which satisfy  $0 < |h| < \delta'$  also satisfy:

$$\left| \frac{|f(a+h) - f(a) - Df(a)(h)|}{|h|} - 0 \right| < \epsilon,$$

which can be rewritten as:

$$|f(a+h) - f(a) - Df(a)(h)| < \epsilon|h|.$$

Then, by the triangle inequality, we obtain:

$$\epsilon|h| > |f(a+h) - f(a) - Df(a)(h)| \geq |f(a+h) - f(a)| - |Df(a)(h)|,$$

which can be rewritten as:

$$|f(a+h) - f(a)| < \epsilon|h| + |Df(a)(h)|.$$

Now, since  $Df(a)$  is a linear map which acts on  $h$ , Assignment 1 Question 2 gives us some  $M \in \mathbb{R}$  such that  $|Df(a)(h)| \leq M|h|$  for all  $h \in \mathbb{R}^n$ . (We can assume without loss of generality that  $M > 0$ ; otherwise, we could pick any  $M' > 0$ , and it would satisfy  $|Df(a)(h)| \leq M|h| \leq M'|h|$  for all  $h \in \mathbb{R}^n$ .) Plugging  $|Df(a)(h)| \leq M|h|$  into the above inequality, we obtain:

$$|f(a+h) - f(a)| < \epsilon|h| + M|h| = (M + \epsilon)|h|. \quad (**)$$

Note that this result is true for all  $h \in \mathbb{R}^n$  which satisfy  $0 < |h| < \delta'$ .

Now, let us define  $\delta := \min(\delta', \frac{\epsilon}{M+\epsilon}) > 0$ . Then, for all  $x \in \mathbb{R}^n$  which satisfy  $0 < |x - a| < \delta$ , we can define  $h := x - a$ , giving us  $0 < |h| < \delta$ . Since  $0 < |h| < \delta \leq \delta'$ , the inequality  $(**)$  applies to  $h$ . We also have  $|h| < \delta \leq \frac{\epsilon}{M+\epsilon}$ . Plugging this into  $(**)$ , we obtain:

$$|f(x) - f(a)| = |f(a+h) - f(a)| < (M + \epsilon)|h| < (M + \epsilon) \frac{\epsilon}{M + \epsilon} = \epsilon.$$

Therefore, for all  $\epsilon > 0$ , we found  $\delta > 0$  such that all  $x \in \mathbb{R}^n$  which satisfy  $|x - a| < \delta$  also satisfy  $|f(x) - f(a)| < \epsilon$ , so  $f$  is continuous at  $a$ , as required.  $\square$

2. (Note: This question was not marked.)

We are given some continuous function  $g : S^1 \rightarrow \mathbb{R}$  such that  $g(0,1) = g(1,0) = 0$  and  $g(-x) = -g(x)$ . We also define  $f : \mathbb{R}^2 \rightarrow \mathbb{R}$  by  $f(0) = 0$  and by  $f(x) = |x|g(\frac{x}{|x|})$  for all nonzero  $x$ .

a) We will prove that given any  $x_0 \in \mathbb{R}^2$ , the function  $h : \mathbb{R} \rightarrow \mathbb{R}$  defined by  $h(t) := f(tx_0)$  is differentiable.

First, for the special case  $x_0 = 0$ , we have  $h(t) = f(t \cdot 0) = 0$  for all  $t \in \mathbb{R}$ , so by Spivak's Theorem 2-3(1),  $h$  is differentiable with  $Dh(a) = 0$  for all  $a \in \mathbb{R}$ .

Now, suppose that  $x_0$  is nonzero. Then, we can scale  $x_0$  to define the point  $y := \frac{x_0}{|x_0|} \in S^1$ . Next, we will prove for the following three cases that  $h(t) = |x_0|g(y)t$ :

**Case 1:**  $t > 0$ . Then, we have  $|tx_0| = |t||x_0| = t|x_0|$ , so:

$$h(t) = f(tx_0) = |tx_0|g\left(\frac{tx_0}{|tx_0|}\right) = t|x_0|g\left(\frac{tx_0}{t|x_0|}\right) = t|x_0|g\left(\frac{x_0}{|x_0|}\right) = |x_0|g(y)t.$$

**Case 2:**  $t = 0$ . Then, we have:

$$h(t) = f(0 \cdot x_0) = f(0) = 0 = |x_0|g(y) \cdot 0 = |x_0|g(y)t.$$

**Case 3:**  $t < 0$ . Then, we have  $|tx_0| = |t||x_0| = -t|x_0|$ , so:

$$h(t) = f(tx_0) = |tx_0|g\left(\frac{tx_0}{|tx_0|}\right) = t|x_0|g\left(\frac{tx_0}{-t|x_0|}\right) = t|x_0|g\left(-\frac{tx_0}{t|x_0|}\right) = t|x_0|g(-y) = |x_0|g(y)t.$$

Therefore, in all three cases, we obtain  $h(t) = |x_0|g(y)t$ . Since  $|x_0|g(y)$  is constant, it follows that  $h$  is a linear map with matrix  $(|x_0|g(y))$ . Then, Spivak's Theorem 2-3(2) states that  $h$  is differentiable, with  $Dh(t) = h$  for all  $t \in \mathbb{R}$ . Overall,  $h$  is differentiable no matter whether  $x_0$  is zero or nonzero, as required.  $\square$

b) We will prove that  $f$  is differentiable at  $(0,0)$  if and only if  $g = 0$ .

First, if  $g = 0$ , then we have  $f(0) = 0$ , as well as  $f(x) = |x|g(\frac{x}{|x|}) = 0$  for all nonzero  $x$ . As a result,  $f$  is the constant function 0. Then, by Spivak's Theorem 2-3(1),  $f$  is differentiable at  $(0,0)$  with  $Df(0,0) = 0$ , as desired.

Next, suppose that  $g$  is not zero everywhere, so there exists some  $z \in S^1$  such that  $g(z) \neq 0$ . Then, assume for contradiction that  $f$  is differentiable at  $(0,0)$ . By definition, this gives us that the error function  $f((0,0) + h) - f(0,0) - Df(0,0)(h)$  is in  $o(h)$ , so:

$$\lim_{h \rightarrow 0} \frac{|f((0,0) + h) - f(0,0) - Df(0,0)(h)|}{|h|} = 0.$$

Applying  $f(0,0) = 0$ , this limit can be rewritten as:

$$\lim_{h \rightarrow 0} \frac{|f(h) - Df(0,0)(h)|}{|h|} = 0.$$

In other words, for all  $\epsilon > 0$ , there exists  $\delta > 0$  such that all  $h \in \mathbb{R}^2$  which satisfy  $0 < |h - 0| < \delta$  also satisfy:

$$\frac{|f(h) - Df(0,0)(h)|}{|h|} = \left| \frac{|f(h) - Df(0,0)(h)|}{|h|} - 0 \right| < \epsilon. \quad (*)$$

Now, consider  $h_1 = (\frac{\delta}{2}, 0) \in \mathbb{R}^2$ . First, we have:

$$0 < |h_1 - 0| = \sqrt{\left(\frac{\delta}{2}\right)^2 + 0^2} = \frac{\delta}{2} < \delta.$$

As a result, we can plug  $h = h_1$  into (\*), so:

$$\begin{aligned} \epsilon &> \frac{\left|f\left(\frac{\delta}{2}, 0\right) - Df(0, 0)\left(\frac{\delta}{2}, 0\right)\right|}{\left|\left(\frac{\delta}{2}, 0\right)\right|} \\ &= \frac{\left|\left(\frac{\delta}{2}, 0\right)\left|g\left(\frac{\left(\frac{\delta}{2}, 0\right)}{\left|\left(\frac{\delta}{2}, 0\right)\right|}\right) - Df(0, 0)\left(\frac{\delta}{2}, 0\right)\right|\right|}{\left|\left(\frac{\delta}{2}, 0\right)\right|} \\ &= \frac{\left|\frac{\delta}{2}g(1, 0) - Df(0, 0)\left(\frac{\delta}{2}, 0\right)\right|}{\frac{\delta}{2}} \\ &= \frac{\left|\frac{\delta}{2}g(1, 0) - \frac{\delta}{2}Df(0, 0)(1, 0)\right|}{\frac{\delta}{2}} && \text{(Since } Df(0, 0) \text{ is a linear map)} \\ &= |g(1, 0) - Df(0, 0)(1, 0)| \\ &= |0 - Df(0, 0)(1, 0)| \\ &= |Df(0, 0)(1, 0)| \end{aligned}$$

Next, consider  $h_2 = (0, \frac{\delta}{2}) \in \mathbb{R}^2$ . First, we have:

$$0 < |h_2 - 0| = \sqrt{0^2 + \left(\frac{\delta}{2}\right)^2} = \frac{\delta}{2} < \delta.$$

As a result, we can plug  $h = h_2$  into (\*), so:

$$\begin{aligned} \epsilon &> \frac{\left|f\left(0, \frac{\delta}{2}\right) - Df(0, 0)\left(0, \frac{\delta}{2}\right)\right|}{\left|\left(0, \frac{\delta}{2}\right)\right|} \\ &= \frac{\left|\left(0, \frac{\delta}{2}\right)\left|g\left(\frac{\left(0, \frac{\delta}{2}\right)}{\left|\left(0, \frac{\delta}{2}\right)\right|}\right) - Df(0, 0)\left(0, \frac{\delta}{2}\right)\right|\right|}{\left|\left(0, \frac{\delta}{2}\right)\right|} \\ &= \frac{\left|\frac{\delta}{2}g(0, 1) - Df(0, 0)\left(0, \frac{\delta}{2}\right)\right|}{\frac{\delta}{2}} \\ &= \frac{\left|\frac{\delta}{2}g(0, 1) - \frac{\delta}{2}Df(0, 0)(0, 1)\right|}{\frac{\delta}{2}} && \text{(Since } Df(0, 0) \text{ is a linear map)} \\ &= |g(0, 1) - Df(0, 0)(0, 1)| \\ &= |0 - Df(0, 0)(0, 1)| \\ &= |Df(0, 0)(0, 1)| \end{aligned}$$

Finally, consider  $h_3 = \frac{\delta}{2}z \in \mathbb{R}^2$ , where  $z \in S^1$  was chosen above such that  $g(z) \neq 0$ . First, we have:

$$0 < |h_3 - 0| = \left| \frac{\delta}{2}z \right| = \left| \frac{\delta}{2} \right| |z| = \frac{\delta}{2} \cdot 1 < \delta.$$

As a result, we can plug  $h = h_3$  into (\*), so:

$$\begin{aligned} \epsilon &> \frac{\left| f\left(\frac{\delta}{2}z\right) - Df(0,0)\left(\frac{\delta}{2}z\right) \right|}{\left| \frac{\delta}{2}z \right|} \\ &= \frac{\left| \frac{\delta}{2}z \left| g\left(\frac{\frac{\delta}{2}z}{\left|\frac{\delta}{2}z\right|}\right) - Df(0,0)\left(\frac{\delta}{2}z\right) \right|}{\left| \frac{\delta}{2}z \right|} \\ &= \frac{\left| \frac{\delta}{2}g\left(\frac{\frac{\delta}{2}z}{\frac{\delta}{2}}\right) - Df(0,0)\left(\frac{\delta}{2}z\right) \right|}{\frac{\delta}{2}} \\ &= \frac{\left| \frac{\delta}{2}g(z) - \frac{\delta}{2}Df(0,0)(z) \right|}{\frac{\delta}{2}} && \text{(Since } Df(0,0) \text{ is a linear map)} \\ &= |g(z) - Df(0,0)(z)| \end{aligned}$$

Overall, we proved that  $|Df(0,0)(1,0)|$  and  $|Df(0,0)(0,1)|$  are both less than  $\epsilon$  for all  $\epsilon > 0$ . This implies that  $|Df(0,0)(1,0)| = |Df(0,0)(0,1)| = 0$ , so  $Df(0,0)(1,0) = Df(0,0)(0,1) = 0$ . Then, since  $(1,0)$  and  $(0,1)$  form a basis for  $\mathbb{R}^2$ , it follows that  $Df(0,0)$  is the zero map. We also proved that  $|g(z) - Df(0,0)(z)| < \epsilon$  for all  $\epsilon > 0$ , which implies  $|g(z) - Df(0,0)(z)| = 0$ , so  $Df(0,0)(z) = g(z)$ . Since  $Df(0,0)$  is the zero map, this contradicts  $g(z) \neq 0$ . Thus, by contradiction,  $f$  is not differentiable at  $(0,0)$  if  $g$  is not zero everywhere.

Overall, we proved that  $f$  is differentiable at  $(0,0)$  if and only if  $g$  is zero everywhere, as required.  $\square$

3. Given any function  $f : \mathbb{R}^n \rightarrow \mathbb{R}$  such that  $|f(x)| \leq |x|^2$ , we will prove that  $f$  is differentiable at 0. In fact, we will prove that the differential of  $f$  at 0 is 0.

Let us define the linear transformation  $L : \mathbb{R}^n \rightarrow \mathbb{R}$  by  $L(h) = 0$  for all  $h \in \mathbb{R}^n$ . Then, to show that  $Df(0) = L$ , we need to show that the error function:

$$e(h) := f(0 + h) - f(0) - L(h) = f(h) - f(0) - L(h)$$

is in  $o(h)$ . First, since  $|f(0)| \leq |0|^2 = 0$ , we obtain  $f(0) = 0$  by the positive definiteness of the norm. We also know that  $L(h) = 0$  for all  $h$ , so:

$$e(h) = f(h) - f(0) - L(h) = f(h) - 0 - 0 = f(h)$$

for all  $h \in \mathbb{R}^n$ . As a result,  $e(0) = f(0) = 0$ .

Now, we wish to show that  $\lim_{h \rightarrow 0} \frac{|e(h)|}{|h|} = 0$ . Let any  $\epsilon > 0$  be given. Then, let us define  $\delta := \epsilon > 0$ . For all  $h \in \mathbb{R}^n$  such that  $0 < |h - 0| < \delta$ , we obtain:

$$\begin{aligned} \left| \frac{|e(h)|}{|h|} - 0 \right| &= \frac{|e(h)|}{|h|} \\ &\leq \frac{|h|^2}{|h|} \\ &= |h| \\ &< \delta \\ &= \epsilon. \end{aligned}$$

Therefore, for all  $\epsilon > 0$ , we found  $\delta > 0$  such that all  $h \in \mathbb{R}^n$  which satisfy  $0 < |h - 0| < \delta$  also satisfy  $\left| \frac{|e(h)|}{|h|} - 0 \right| < \epsilon$ , so  $\lim_{h \rightarrow 0} \frac{|e(h)|}{|h|} = 0$ . This, combined with  $e(0) = 0$ , proves that  $e(h)$  is in  $o(h)$ . Therefore, it follows that  $f$  is differentiable at 0 with  $Df(0) = L = 0$ , as required.  $\square$

4. We will evaluate  $f'$  for the following functions:

a)  $f(x, y, z) := x^y$ .

First, let  $f_1 : \mathbb{R}^3 \rightarrow \mathbb{R}^2$ ,  $f_2 : \mathbb{R}^2 \rightarrow \mathbb{R}$ , and  $f_3 : \mathbb{R} \rightarrow \mathbb{R}$  be defined by  $f_1(x, y, z) := (\log(x), y)$ ,  $f_2(x, y) := xy$ , and  $f_3(x) := e^x$ . Then, we can write  $f$  as the following composition:

$$f(x, y, z) = x^y = (e^{\log(x)})^y = e^{\log(x)y} = f_3(f_2(f_1(x, y, z))) = (f_3 \circ f_2 \circ f_1)(x, y, z).$$

Now, from MAT157, we know that  $f_3'(a) = (e^a)$  for all  $a \in \mathbb{R}$ . As a result:

$$f_3'(f_2(f_1(x, y, z))) = f_3'(f_2(\log(x), y)) = f_3'(\log(x)y) = (e^{\log(x)y}) = (x^y).$$

Next, from Spivak's Theorem 2-3(5), we know that  $f_2'(a, b) = (b, a)$  for all  $(a, b) \in \mathbb{R}^2$ , so:

$$f_2'(f_1(x, y, z)) = f_2'(\log(x), y) = (y, \log(x)).$$

Finally, we wish to evaluate  $f_1'(x, y, z)$ . First, we can decompose  $f_1$  further into  $f_1 = (g_1, g_2)$ , where the functions  $g_1, g_2 : \mathbb{R}^3 \rightarrow \mathbb{R}$  are defined by  $g_1(x, y, z) := \log(x)$  and  $g_2(x, y, z) := y$ . To evaluate  $g_1'(x, y, z)$ , we write  $g_1$  as the composition  $h_2 \circ h_1$ , where  $h_1 : \mathbb{R}^3 \rightarrow \mathbb{R}$  is defined by  $h_1(x, y, z) := x$ , and  $h_2 : \mathbb{R} \rightarrow \mathbb{R}$  is defined by  $h_2(x) := \log(x)$ . Finally, we can find these derivatives. Since  $h_1$  is the linear transformation with matrix  $(1, 0, 0)$ , Spivak's Theorem 2-3(2) tells us that  $h_1'(x, y, z) = (1, 0, 0)$ , and MAT157 also tells us that  $h_2'(x) = \frac{1}{x}$ . Thus, the Chain Rule gives us:

$$g_1'(x, y, z) = h_2'(h_1(x, y, z)) \cdot h_1'(x, y, z) = h_2'(x) \cdot (1, 0, 0) = \left(\frac{1}{x}\right) \cdot (1, 0, 0) = \left(\frac{1}{x}, 0, 0\right).$$

Meanwhile, since  $g_2$  is a linear transformation with the matrix  $(0, 1, 0)$ , Spivak's Theorem 2-3(2) tells us that  $g_2'(x, y, z) = (0, 1, 0)$ . Overall, applying Spivak's Theorem 2-3(3), we obtain:

$$f_1'(x, y, z) = \begin{pmatrix} g_1'(x, y, z) \\ g_2'(x, y, z) \end{pmatrix} = \begin{pmatrix} \frac{1}{x} & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix}.$$

Finally, applying the Chain Rule twice on  $f = f_3 \circ f_2 \circ f_1$ , we obtain:

$$\begin{aligned} f'(x, y, z) &= (f_3 \circ f_2)'(f_1(x, y, z)) \cdot f_1'(x, y, z) \\ &= f_3'(f_2(f_1(x, y, z))) \cdot f_2'(f_1(x, y, z)) \cdot f_1'(x, y, z) \\ &= (x^y) \cdot (y, \log(x)) \cdot \begin{pmatrix} \frac{1}{x} & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix} \\ &= (x^y) \cdot \left(\frac{y}{x}, \log(x), 0\right) \\ &= \boxed{(x^{y-1}y, x^y \log(x), 0)} \end{aligned}$$

□

b)  $f(x, y, z) := (x^y, z)$ .

First, we decompose  $f$  into  $f = (f_1, f_2)$ , where  $f_1 : \mathbb{R}^3 \rightarrow \mathbb{R}$  is defined by  $f_1(x, y, z) := x^y$  and  $f_2 : \mathbb{R}^3 \rightarrow \mathbb{R}$  is defined by  $f_2(x, y, z) := z$ . Then, from part a),  $f_1'(x, y, z) = (x^{y-1}y, x^y \log(x), 0)$ . Moreover,  $f_2$  is a linear transformation with the matrix  $(0, 0, 1)$ , so we know from Spivak's Theorem 2-3(2) that  $f_2'(x, y, z) = (0, 0, 1)$ . Therefore, applying Spivak's Theorem 2-3(3), we obtain:

$$f'(x, y, z) = \begin{pmatrix} f_1'(x, y, z) \\ f_2'(x, y, z) \end{pmatrix} = \boxed{\begin{pmatrix} x^{y-1}y & x^y \log(x) & 0 \\ 0 & 0 & 1 \end{pmatrix}}.$$

□

c)  $f(x, y, z) := (x + y)^z$ .

First, let us define the functions  $f_1 : \mathbb{R}^3 \rightarrow \mathbb{R}^3$  and  $f_2 : \mathbb{R}^3 \rightarrow \mathbb{R}$  by  $f_1(x, y, z) := (x + y, z, 0)$  and  $f_2(x, y, z) := x^y$ . Then, we can write  $f(x, y, z)$  as the following composition:

$$f(x, y, z) = (x + y)^z = f_2(x + y, z, 0) = (f_2 \circ f_1)(x, y, z).$$

Now, from part a), we know that  $f'_2(x, y, z) = (x^{y-1}y, x^y \log(x), 0)$ , so we obtain:

$$f'_2(f_1(x, y, z)) = f'_2(x + y, z, 0) = ((x + y)^{z-1}z, (x + y)^z \log(x + y), 0).$$

Additionally, since  $f_1$  is a linear transformation with the matrix  $\begin{pmatrix} 1 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix}$ , Spivak's Theorem

2-3(2) tells us that  $f'_1(x, y, z) = \begin{pmatrix} 1 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix}$ . Therefore, applying the Chain Rule on  $f = f_2 \circ f_1$ ,

we obtain:

$$\begin{aligned} f'(x, y, z) &= f'_2(f_1(x, y, z)) \cdot f'_1(x, y, z) \\ &= ((x + y)^{z-1}z, (x + y)^z \log(x + y), 0) \cdot \begin{pmatrix} 1 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix} \\ &= \boxed{((x + y)^{z-1}z, (x + y)^{z-1}z, (x + y)^z \log(x + y))}. \end{aligned}$$

□



5. a) We will prove that if  $f : \mathbb{R}^n \times \mathbb{R}^m \rightarrow \mathbb{R}^p$  is bilinear, then:

$$\lim_{(h,k) \rightarrow 0} \frac{|f(h,k)|}{|(h,k)|} = 0.$$

First, for all  $1 \leq i \leq n$ , let  $e_i$  be the point in  $\mathbb{R}^n$  with its  $i^{\text{th}}$  coordinate equal to 1 and all other coordinates equal to 0. Then, let  $g_i : \mathbb{R}^m \rightarrow \mathbb{R}^p$  be the linear map defined by  $g_i(k) := f(e_i, k)$  for all  $k \in \mathbb{R}^m$  – this map is linear because  $e_i$  is constant. Now, in Assignment 1 Question 2, we proved that there exists some  $M_i \in \mathbb{R}$  such that  $|g_i(k)| \leq M_i|k|$  for all  $k \in \mathbb{R}^m$ . We can assume without loss of generality that  $M_i > 0$ ; otherwise, we could pick any  $M'_i > 0$ , and it would satisfy  $|g_i(k)| \leq M_i|k| \leq M'_i|k|$ .

Next, for all  $h = (h_1, \dots, h_n) \in \mathbb{R}^n$ , we have  $h = h_1e_1 + \dots + h_n e_n$ . Then, since  $f$  is bilinear, we obtain:

$$\begin{aligned} |f(h,k)| &= |f(h_1e_1 + \dots + h_n e_n, k)| \\ &= |h_1f(e_1, k) + \dots + h_n f(e_n, k)| \\ &\leq |h_1f(e_1, k)| + \dots + |h_n f(e_n, k)| && \text{(Applying triangle inequality)} \\ &= |h_1||f(e_1, k)| + \dots + |h_n||f(e_n, k)| \\ &= |h_1||g_1(k)| + \dots + |h_n||g_n(k)| \\ &\leq |h_1|M_1|k| + \dots + |h_n|M_n|k| \\ &= (M_1|h_1| + \dots + M_n|h_n|)|k| \end{aligned}$$

Now, let us define  $M := \max(M_1, \dots, M_n) > 0$ . Then, we obtain:

$$|f(h,k)| \leq (M_1|h_1| + \dots + M_n|h_n|)|k| \leq M(|h_1| + \dots + |h_n|)|k|.$$

Next, let us define the point  $z = (z_1, \dots, z_n) \in \mathbb{R}^n$  by  $z_i = -1$  if  $h_i$  is negative, and  $z_i = 1$  otherwise. In other words,  $z_i h_i = |h_i|$ . Then, by the Cauchy-Schwarz inequality, we obtain:

$$\begin{aligned} \langle h, z \rangle &\leq |h| \cdot |z| \\ h_1 z_1 + \dots + h_n z_n &\leq |h| \cdot \sqrt{z_1^2 + \dots + z_n^2} \\ |h_1| + \dots + |h_n| &\leq |h| \cdot \sqrt{(\pm 1)^2 + \dots + (\pm 1)^2} \\ |h_1| + \dots + |h_n| &\leq \sqrt{n}|h| \end{aligned}$$

As a result, we obtain:

$$|f(h,k)| \leq M(|h_1| + \dots + |h_n|)|k| \leq M\sqrt{n}|h||k|.$$

Now, since  $(|h| - |k|)^2 \geq 0$ , we have  $|h|^2 - 2|h||k| + |k|^2 \geq 0$ , so:

$$|h||k| \leq \frac{1}{2}(|h|^2 + |k|^2) = \frac{1}{2}|(h,k)|^2.$$

Finally, we obtain:

$$\frac{|f(h,k)|}{|(h,k)|} \leq \frac{M\sqrt{n}|h||k|}{|(h,k)|} \leq \frac{M\sqrt{n}}{2}|(h,k)|$$

for all nonzero  $(h, k)$ .

Now, given any  $\epsilon > 0$ , let us define  $\delta := \frac{2\epsilon}{M\sqrt{n}} > 0$ . Then, for all  $(h, k) \in \mathbb{R}^n \times \mathbb{R}^m$  satisfying  $0 < |(h, k) - 0| < \delta$ , we obtain:

$$\left| \frac{|f(h, k)|}{|(h, k)|} - 0 \right| = \frac{|f(h, k)|}{|(h, k)|} \leq \frac{M\sqrt{n}}{2} |(h, k)| < \frac{M\sqrt{n}}{2} \delta = \epsilon.$$

Therefore, for all  $\epsilon > 0$ , we found  $\delta > 0$  such that all  $(h, k)$  which satisfy  $0 < |(h, k) - 0| < \delta$  also satisfy  $\left| \frac{|f(h, k)|}{|(h, k)|} - 0 \right| < \epsilon$ , so:

$$\lim_{(h, k) \rightarrow 0} \frac{|f(h, k)|}{|(h, k)|} = 0,$$

as required. □

b) We will prove that  $Df(a, b)(x, y) = f(a, y) + f(x, b)$  whenever  $f$  is bilinear. In particular, for all points  $(a, b) \in \mathbb{R}^n \times \mathbb{R}^m$ , let us define the linear map  $L(a, b) : \mathbb{R}^n \times \mathbb{R}^m \rightarrow \mathbb{R}^p$  by  $L(a, b)(x, y) := f(a, y) + f(x, b)$ . (This map is linear if we treat  $a$  and  $b$  as constants because  $f$  is bilinear.) Then, by definition of differentiation, we need to show that:

$$e(h) := f((a, b) + h) - f(a, b) - L(a, b)(h) \in o(h).$$

First, if we decompose  $h$  as  $h = (h_1, h_2)$ , where  $h_1 \in \mathbb{R}^n$  and  $h_2 \in \mathbb{R}^m$ , then we can use the fact that  $f$  is bilinear multiple times to obtain:

$$\begin{aligned} e(h_1, h_2) &= f(a + h_1, b + h_2) - f(a, b) - L(a, b)(h_1, h_2) \\ &= f(a + h_1, b + h_2) - f(a, b) - f(a, h_2) - f(h_1, b) \quad (\text{Applying definition of } L(a, b)) \\ &= (f(a, b + h_2) + f(h_1, b + h_2)) - f(a, b) - f(a, h_2) - f(h_1, b) \\ &= f(a, b + h_2) + f(h_1, b + h_2) - f(a, b + h_2) - f(h_1, b) \\ &= f(h_1, b + h_2) - f(h_1, b) \\ &= f(h_1, (b + h_2) - b) \\ &= f(h_1, h_2) \end{aligned}$$

Now, if we plug in  $h = (h_1, h_2) = 0$ , since  $f$  is bilinear, we obtain:

$$e(0) = f(0, 0) = f(0 \cdot 1, 0) = 0 \cdot f(1, 0) = 0.$$

Moreover, since  $f$  is bilinear, part a) gives us that:

$$\lim_{(h_1, h_2) \rightarrow 0} \frac{|e(h_1, h_2)|}{|(h_1, h_2)|} = \lim_{(h_1, h_2) \rightarrow 0} \frac{|f(h_1, h_2)|}{|(h_1, h_2)|} = 0.$$

These two pieces of information prove that  $e(h) \in o(h)$ . Therefore, we obtain:

$$Df(a, b)(x, y) = L(a, b)(x, y) = f(a, y) + f(x, b),$$

as required. □

c) We will prove that the formula  $Dp(a, b)(x, y) = bx + ay$ , where  $p : \mathbb{R} \times \mathbb{R} \rightarrow \mathbb{R}$  is defined by  $p(x, y) = xy$ , is a special case of part b).

First, we will show that  $p$  is bilinear. If  $y$  is kept constant, then  $p(x, y)$  is a linear transformation in

terms of  $x$  with matrix  $(y)$ . Additionally, if  $x$  is kept constant, then  $p(x, y)$  is a linear transformation in terms of  $y$  with matrix  $(x)$ . Since  $p$  is linear in terms of each variable when the other variable is kept constant,  $p$  is bilinear, as desired.

Now, when we apply the formula in part b), we obtain:

$$Dp(a, b)(x, y) = p(a, y) + p(x, b) = ay + xb = bx + ay.$$

Therefore, the formula  $Dp(a, b)(x, y) = bx + ay$  is a special case of part b), as required.  $\square$

6. (Note: This question was not marked.)

We will prove that the function  $f : \mathbb{R}^4 \rightarrow \mathbb{R}$  defined by  $f(a, b, c, d) := \det \begin{pmatrix} a & b \\ c & d \end{pmatrix} = ad - bc$  is

differentiable, and we will also compute its differential.

First, let us define the functions  $f_1 : \mathbb{R}^4 \rightarrow \mathbb{R}^2$  and  $f_2 : \mathbb{R}^2 \rightarrow \mathbb{R}$  by  $f_1(a, b, c, d) := (ad, bc)$  and  $f_2(a, b) := a - b$ . Then, we can write  $f$  as the composition  $f = f_2 \circ f_1$ . Now, let us decompose  $f_1$  further into  $f_1 = (g_1, g_2)$ , where  $g_1 : \mathbb{R}^4 \rightarrow \mathbb{R}$  is defined by  $g_1(a, b, c, d) := ad$  and  $g_2 : \mathbb{R}^4 \rightarrow \mathbb{R}$  is defined by  $g_2(a, b, c, d) := bc$ . Then,  $g_1$  can be written as a composition  $g_1 = h_2 \circ h_1$ , where  $h_1 : \mathbb{R}^4 \rightarrow \mathbb{R}^2$  is defined by  $h_1(a, b, c, d) = (a, d)$  and  $h_2 : \mathbb{R}^2 \rightarrow \mathbb{R}$  is defined by  $h_2(a, b) = ab$ .

Since  $h_1$  is a linear transformation with matrix  $\begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$ , Spivak's Theorem 2-3(2) gives us

that  $h_1$  is differentiable and that  $h_1'(a, b, c, d) = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$ . Additionally, Spivak's Theorem 2-3(5) states that  $h_2$  is also differentiable and that  $h_2'(a, b) = (b, a)$ . Thus, applying the Chain Rule,  $g_1 = h_2 \circ h_1$  is differentiable, and:

$$\begin{aligned} g_1'(a, b, c, d) &= h_2'(h_1(a, b, c, d)) \cdot h_1'(a, b, c, d) \\ &= h_2'(a, d) \cdot h_1'(a, b, c, d) \\ &= (d, a) \cdot \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix} \\ &= (d, 0, 0, a). \end{aligned}$$

Similarly,  $g_2$  can also be written as a composition  $g_2 = h_2 \circ h_3$ , where  $h_3 : \mathbb{R}^4 \rightarrow \mathbb{R}^2$  is defined by  $h_3(a, b, c, d) = (b, c)$ . Since  $h_3$  is a linear transformation with matrix  $\begin{pmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{pmatrix}$ , Spivak's

Theorem 2-3(2) gives us that  $h_3$  is differentiable and that  $h_3'(a, b, c, d) = \begin{pmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{pmatrix}$ . Thus, applying the Chain Rule,  $g_2 = h_2 \circ h_3$  is differentiable, and:

$$\begin{aligned} g_2'(a, b, c, d) &= h_2'(h_3(a, b, c, d)) \cdot h_3'(a, b, c, d) \\ &= h_2'(b, c) \cdot h_3'(a, b, c, d) \\ &= (c, b) \cdot \begin{pmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{pmatrix} \\ &= (0, c, b, 0). \end{aligned}$$

Now, by Spivak's Theorem 2-3(3), the function  $f_1 = (g_1, g_2)$  is differentiable, and:

$$f_1'(a, b, c, d) = \begin{pmatrix} g_1'(a, b, c, d) \\ g_2'(a, b, c, d) \end{pmatrix} = \begin{pmatrix} d & 0 & 0 & a \\ 0 & c & b & 0 \end{pmatrix}.$$

Next, since the function  $f_2(a, b) = a - b$  is a linear transformation with matrix  $(1, -1)$ , Spivak's Theorem 2-3(2) gives us that  $f_2$  is differentiable and that  $f_2'(a, b) = (1, -1)$ . Finally, applying

the Chain Rule, the function  $f = f_2 \circ f_1$  is differentiable, and we obtain:

$$\begin{aligned} f'(a, b, c, d) &= f_2'(f_1(a, b, c, d)) \cdot f_1'(a, b, c, d) \\ &= (1, -1) \cdot \begin{pmatrix} d & 0 & 0 & a \\ 0 & c & b & 0 \end{pmatrix} \\ &= \boxed{(d, -c, -b, a)}. \end{aligned}$$

□

## Notes on Intuition

Now, let's develop some intuition on how to approach these problems and find solutions for them. (Note: This section was not submitted on Crowdmark.)

1. We wish to prove that  $f$  is continuous at  $a$ . Since  $f$  is diffable at  $a$ , we know that  $f(a+h) - f(a)$  equals a linear correction  $L(h)$ , plus a "tiny" error function  $e(h)$ . As a result, we can model our solution after A2 Q4 (where we proved that linear transformations are continuous). Due to the "tiny" error function, our solution also required some modifications where we place a linear bound (such as  $\epsilon|h|$ ) on the "tiny" error function.
2. a) First, we will provide an intuitive explanation of the formula  $f(x) := |x|g(\frac{x}{|x|})$ . The expression  $\frac{x}{|x|}$  rescales  $x$  to project it onto the unit circle  $S^1$  so that we can input it into  $g$ . Then, we multiply the output by  $|x|$  to scale it according to the size of  $|x|$ . Now, if we restrict  $f$  to any line through the origin, then the projection of this line onto  $S^1$  consists of two diametrically opposite points,  $y$  and  $-y$ . This gives us lots of consistency as we input these projections into  $g$ . Then,  $f(tx)$  is proportional to  $\pm|tx|$ , which is itself proportional to  $\pm|t|$ . Overall, these observations motivate us to show that  $h$  is linear, from which it follows that  $h$  is differentiable.  
b) When we consider all points  $h$  arbitrarily close to  $(0,0)$ , based on part a), we can learn about  $f(h)$  by inputting the projection of  $h$  into  $g$ . These points  $h$  can get projected onto any point on  $S^1$ . Based on the textbook's hint, if we choose  $h$  which get projected horizontally or vertically onto  $S^1$ , then  $g$  will output 0, so  $f$  will also output 0. Due to linear algebra, this constrains  $Df(0,0)$  to be zero. However, if we choose  $h$  which get projected onto a point  $z \in S^1$  such that  $g(z) \neq 0$ , then  $f$  becomes a nonzero linear map when restricted along the direction of  $z$ . This contradicts  $Df(0,0) = 0$ , which allows us to complete our proof by contradiction.
3. The main idea for this problem is that if  $|f(x)|$  is bounded by  $|x|^2$ , where  $|x|^2$  becomes "tiny" as  $x$  approaches 0, then  $f(x)$  itself should be tiny. Then, if we want to approximate  $f(x)$  as a linear map where  $x$  is near 0, we should choose the zero map. Another way to see this is that  $|x|^2$  has a parabolic shape near  $x = 0$ , so  $|x|^2$  is flat at  $x = 0$ , and then  $f(x)$  itself is flat at  $x = 0$  since it is bounded by  $|x|^2$ . This motivates us to choose the linear map  $L$  defined by  $L(h) = 0$ . After considering the error function  $e(h)$  and performing some computations to determine that  $\frac{|e(h)|}{|h|} \leq |h|$ , this motivates us to select  $\delta = \epsilon$  in our delta-epsilon proof so that  $|h| < \delta = \epsilon$ .
4. a) Recall that, in lecture, we computed the differential of the function  $g(x, y) = \frac{x}{y}$  by writing it in the exponential form  $g(x, y) = e^{\log(x) - \log(y)}$ . Our solution for  $f(x, y, z) = x^y$  is heavily motivated by this proof technique. After writing  $f$  in the exponential form  $f(x, y, z) = e^{\log(x)y}$ , we finish the problem using the Chain Rule, very similarly to the computation of  $g'$  done in class.  
b) We immediately notice that the output of the function  $f(x, y, z) = (x^y, z)$  contains  $x^y$ , whose differential we found in part a), and  $z$ , whose differential is relatively easy to find. This motivates us to decompose the output coordinate-wise, then combine the differentials of the coordinates  $x^y$  and  $z$  using Spivak's Theorem 2-3(3).  
c) We wish to use part a) to solve part c) directly, except the base and exponent of  $(x+y)^z$  are slightly different. We account for this modification with a linear transformation of the coordinates. Then, we can finish this problem with the Chain Rule.
5. a) Since  $f(h, k)$  is linear in terms of  $h$ , we expect  $f(h, k)$  to be bounded proportionally to  $|h|$  because of A1 Q2. By the same reasoning,  $f(h, k)$  should also be bounded proportionally to  $k$ .

Overall, we wish to show that  $f(h, k)$  is bounded proportionally to  $|h||k|$ . This product is of degree 2, so it becomes "tiny" as  $(h, k)$  approaches 0, which will help us show that  $f$  is "tiny".

Now, let us figure out how to prove that  $f(h, k)$  is bounded proportionally to  $|h||k|$ . By defining the maps  $g_i(k) := f(e_i, k)$ , we use one of the components of  $f$ 's bilinearity (i.e., the fact that  $f$  is linear in terms of  $k$ ) to bound  $f$  proportionally to  $|k|$ . Next, we need to bound  $f$  proportionally to  $|h|$  as well. To do this, we model our solution after A1 Q2 – note how the key Cauchy-Schwarz application reappears. As discussed above, once  $f(h, k)$  is bounded proportionally to  $|h||k|$ , we are done.

b) The problem already gives us the candidate  $Df(a, b)(x, y) = f(a, y) + f(x, b)$  for the differential of  $f$ , so we only need to verify that it satisfies the definition for differentials. When we work to verify this, the bilinearity of  $f$  allows us to make useful simplifications. Finally, when we compute that the error function is  $f$  itself, part a) (along with some technical details) confirms for us that  $f$  is "tiny", as desired.

c) This sub-problem was relatively straightforward with a couple of formal steps: Verify that  $p(x, y) = xy$  is bilinear (so that we can apply part b)), then perform some computations after applying part b).

6. We observe that the determinant,  $ad - bc$ , is ultimately a difference of products. Then, this problem becomes relatively straightforward because we know how to differentiate products (e.g., by Spivak's Theorem 2-3(5)), and we also know how to differentiate differences.