

## PROBLEM SET #08- ANSWER KEY

## SECOND CALCULUS PROBLEM SET

(1) Consider the function  $f(x, y, z) = xyz$ , with  $y = x^2$  and  $z = x^{1/3}$ .

(a) Rewrite  $f$  as a function  $g : \mathbb{R} \rightarrow \mathbb{R}$  alone and compute  $g'(\cdot)$ . Using  $g'$ , approximate the change in  $f$  when  $x$  increases by 0.1 units, starting from  $(8, 64, 2)$ .

**Ans:**

$$\begin{aligned} g(x) &= f(x, x^2, x^{1/3}) = x \times x^2 \times x^{1/3} \\ &= x^{10/3} \quad \text{so that} \quad g'(x) = 10/3 x^{7/3} \\ g'(8) &= 10/3 \times 8^{7/3} = 10/3 * 2^7 = 426.6667 \\ dg &= df = 0.1 \times 10/3 \times 128 = 42.6. \end{aligned}$$

(b) Compute the total derivative of  $f$  with respect to  $x$ . Using the total derivative, approximate the change in  $f$  when  $x$  increases by 0.1 units, starting from  $(8, 64, 2)$ .

**Ans:**

$$\begin{aligned} \frac{df}{dx} &= f_x + f_y \frac{dy}{dx} + f_z \frac{dz}{dx} = yz + xz \times 2x + yx \times 1/3 x^{-2/3} \\ &= x^{7/3} + x^{4/3} \times 2x + x^3 \times 1/3 x^{-2/3} \\ &= x^{7/3} (1 + 2 + 1/3) = 10/3 x^{7/3} \\ \text{so that} \quad df &= \frac{df(8)}{dx} = 0.1 \times 10/3 \times 8^{7/3} = 42.6 \end{aligned}$$

(c) Write down the differential of  $f$  at  $(8, 64, 2)$ . Using the differential, approximate the change in  $f$  when  $x$  increases by 0.1 units, starting from  $(8, 64, 2)$ .

**Ans:**

$$\begin{aligned}\nabla f(x, y, z) &= [yz \quad xz \quad yx] \quad \text{so that} \quad \nabla f(8, 64, 2) = [128 \quad 16 \quad 512] \\ df &= \nabla f(x, y, z) \begin{bmatrix} dx \\ y'(x)dx \\ z'(x)dx \end{bmatrix}' = [128 \quad 16 \quad 512] \begin{bmatrix} 0.1 \\ 16 \times 0.1 \\ 0.0833 \times 0.1 \end{bmatrix} \\ &= 12.8 + 16 \times 0.16 + 512 \times 0.025 = 42.7\end{aligned}$$

- (d) Identify the direction  $h^*$  that  $(x, y, z)$  moves in, starting from  $(8, 64, 2)$ , when  $x$  increases. Write down the directional derivative of  $f$  in the direction  $\mathbf{h}^*$ , i.e.,  $f_{\mathbf{h}^*}(\cdot, \cdot, \cdot)$ , and evaluate this derivative at  $(8, 64, 2)$ . Using  $f_{\mathbf{h}^*}(8, 64, 2)$ , approximate the change in  $f$  when  $x$  increases by 0.1 units, starting from  $(8, 64, 2)$ .

**Ans:** When  $x$  increases by one, the vector  $(x, y, z)$  increases in the direction  $(dx, y'(x)dx, z'(x)dx)$ . When  $x = 8$  and  $dx = 1$ , therefore,  $(x, y, z)$  increases in the direction  $h^* = (1, 16, 0.0833)$ . The *unit length* vector pointing in this direction is  $h^*/\|h^*\| = (0.0624, 0.9980, 0.0052)$ . Therefore, using the differential to compute the directional derivative,  $f_{\mathbf{h}^*}(8, 64, 2) = [128 \quad 16 \quad 512] \begin{bmatrix} 0.0624 \\ 0.9980 \\ 0.0052 \end{bmatrix} = 26.613$ . Now a  $dx$  of 0.1 induces a shift in  $\mathbb{R}^3$  of  $0.1h^*$ , which has length  $0.1\|h^*\|$ .  $df = f_{\mathbf{h}^*}(8, 64, 2) \times 0.1\|h^*\| = 42.67$

- (e) Check to see that all four of these distinct methods give you the same answer!

**Ans:** Amazingly, they do!

- (2) Recall that a function  $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$  is nothing more than  $m$  functions,  $f^1 \dots f^m$ , each mapping  $\mathbb{R}^n \rightarrow \mathbb{R}$ , and stacked on top of each other.

- (a) Using this fact, write down a formal definition of the directional derivative of  $f$  at  $x_0$  in the direction  $\mathbf{h} \in \mathbb{R}^n$ , for a function  $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$ . Your definition should be of the

form

$$\text{blah, blah} = \lim_{k \rightarrow \infty} \frac{\text{blah} \quad \text{blah}}{\text{blah, blah}}$$

**Ans:** Definition: Given  $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$  and  $\mathbf{h} \in \mathbb{R}^n$ , the directional derivative of  $f$  at  $\mathbf{x}_0$  in the direction  $\mathbf{h}$  is given by, for  $i = 1, \dots, m$ ,

$$f_{\mathbf{h}}^i = \lim_{|k| \rightarrow \infty} \frac{(f^i(\mathbf{x}_0 + \mathbf{h}/k) - f^i(\mathbf{x}_0))}{\|\mathbf{h}\|/k}$$

- (b) Consider the function  $f : \mathbb{R}^2 \rightarrow \mathbb{R}^2$  defined by, for  $i = 1, 2$ ,  $f^i(x, y) = x^{i/3}y^{1-i/3}$ . Using the formal definition in (a) above, compute  $f_{\mathbf{h}^*}(27, 8)$ , where  $\mathbf{h}^* = (54, 16)$ . (Hint:  $(27, 8) + (54, 16)/k = (27, 8)(1 + 2/k)$ ).

**Ans:**

$$\begin{aligned} f_{\mathbf{h}^*}^i(27, 8) &= \lim_{|k| \rightarrow \infty} \frac{(f^i((27, 8) + (54, 16)/k) - f^i(27, 8))}{\|\mathbf{h}\|/k} \\ &= \lim_{|k| \rightarrow \infty} \frac{(f^i((27, 8)(1 + 2/k)) - f^i(27, 8))}{\|\mathbf{h}\|/k}. \end{aligned} \quad \text{Therefore}$$

$$f_{\mathbf{h}^*}^1(27, 8) = \lim_{|k| \rightarrow \infty} \frac{((3 \times 4)(1 + 2/k)) - (3 \times 4)}{\|\mathbf{h}\|/k} = \frac{24}{\|\mathbf{h}\|}. \quad \text{Similarly}$$

$$f_{\mathbf{h}^*}^2(27, 8) = \lim_{|k| \rightarrow \infty} \frac{((9 \times 2)(1 + 2/k)) - (9 \times 2)}{\|\mathbf{h}\|/k} = \frac{36}{\|\mathbf{h}\|} \quad \text{Therefore}$$

$$f_{\mathbf{h}^*}(27, 8) = \begin{bmatrix} 0.4261 \\ 0.6392 \end{bmatrix}$$

- (c) Now compute  $f_{\mathbf{h}^*}(27, 8)$  using the differential of  $f$  at  $(27, 8)$ .

**Ans:**

$$\begin{aligned} \nabla f^i(x, y) &= \left[ \frac{if^i(x, y)}{3x} \quad \frac{(3-i)f^i(x, y)}{3y} \right] \\ f_{\mathbf{h}^*}^i(27, 8) &= \left[ \frac{if^i(x, y)}{3x} \quad \frac{(3-i)f^i(x, y)}{3y} \right] \begin{bmatrix} 54/\|\mathbf{h}^*\| \\ 16/\|\mathbf{h}^*\| \end{bmatrix} && \text{Therefore} \\ \begin{bmatrix} f_{\mathbf{h}^*}^1(27, 8) \\ f_{\mathbf{h}^*}^2(27, 8) \end{bmatrix} &= \begin{bmatrix} \left[ \frac{12}{3 \times 27} \quad \frac{2 \times 12}{3 \times 8} \right] \begin{bmatrix} 54/\|\mathbf{h}^*\| \\ 16/\|\mathbf{h}^*\| \end{bmatrix} \\ \left[ \frac{2 \times 18}{3 \times 27} \quad \frac{18}{3 \times 8} \right] \begin{bmatrix} 54/\|\mathbf{h}^*\| \\ 16/\|\mathbf{h}^*\| \end{bmatrix} \end{bmatrix} \\ &= \frac{1}{\|\mathbf{h}^*\|} \begin{bmatrix} 8 + 16 \\ 24 + 12 \end{bmatrix} = \begin{bmatrix} 0.4261 \\ 0.6392 \end{bmatrix} \end{aligned}$$

(d) Check to see that all of these three distinct methods give you the same answer!

**Ans:** Amazingly, they do!

(3) Consider the function  $f(\mathbf{x}) = x_1^\rho + x_2^\rho$ , where  $\rho \in (-\infty, 1]$ . The whole point here is to use the differential of  $\nabla f$  to answer the following questions, i.e., to answer all parts of the question, approximate  $\nabla f(\mathbf{x} + \mathbf{h}) - \nabla f(\mathbf{x})$  using the differential of  $\nabla f$  at  $\mathbf{x}$ , evaluated at  $\mathbf{h}$ . There are lots of other ways to answer these questions, but the purpose of this question is to give you practice in using the differential of a vector-valued function.

- (a) Check that, up to a first order approximation,<sup>1</sup>  $f$  is homothetic (cf the notes for lecture CALCULUS3,<sup>2</sup> specifically the second example in the subsection entitled Four Graphical Examples.<sup>3</sup>)

**Ans:** The following argument is not entirely rigorous, but can be made so. We need to show that if  $\mathbf{x}$  and  $d\mathbf{x}$  are colinear, (i.e., if  $d\mathbf{x} = \lambda\mathbf{x}$ ), then  $\nabla f(\mathbf{x})$  and  $\nabla f(\mathbf{x} + d\mathbf{x})$  are colinear, i.e.,  $\nabla f(\mathbf{x} + d\mathbf{x}) = \delta \nabla f(\mathbf{x})$ , for some  $\delta > 0$ .

Now  $\nabla f(\mathbf{x}) = [\rho x_1^{\rho-1} \quad \rho x_2^{\rho-1}]$  so that  $Hf(\mathbf{x}) = \begin{bmatrix} \rho(\rho-1)x_1^{\rho-2} & 0 \\ 0 & \rho(\rho-1)x_2^{\rho-2} \end{bmatrix}$  Therefore, for  $d\mathbf{x} = \lambda\mathbf{x}$ ,

$$\begin{aligned} \nabla f(\mathbf{x} + d\mathbf{x}) &\approx \nabla f(\mathbf{x}) + \begin{bmatrix} \rho(\rho-1)x_1^{\rho-2} & 0 \\ 0 & \rho(\rho-1)x_2^{\rho-2} \end{bmatrix} \begin{bmatrix} dx_1 \\ dx_2 \end{bmatrix} \\ &= \begin{bmatrix} \rho x_1^{\rho-1} \\ \rho x_2^{\rho-1} \end{bmatrix} + \begin{bmatrix} \rho(\rho-1)x_1^{\rho-2} & 0 \\ 0 & \rho(\rho-1)x_2^{\rho-2} \end{bmatrix} \begin{bmatrix} \lambda x_1 \\ \lambda x_2 \end{bmatrix} \\ &= \begin{bmatrix} \rho x_1^{\rho-1} \\ \rho x_2^{\rho-1} \end{bmatrix} + \begin{bmatrix} \lambda \rho(\rho-1)x_1^{\rho-1} \\ \lambda \rho(\rho-1)x_2^{\rho-1} \end{bmatrix} = \rho(1 + \lambda(\rho-1)) \begin{bmatrix} x_1^{\rho-1} \\ x_2^{\rho-1} \end{bmatrix} \\ &= (1 + \lambda(\rho-1)) \nabla f(\mathbf{x}) \end{aligned}$$

The argument is not entirely rigorous because of the approximation relationship above. To make it fully rigorous, we could use the 2-dimensional analog of the Taylor Lagrange theorem.

<sup>1</sup> The qualifier “up to a first order approximation” means: you should *pretend* that the answer you get using the differential is exactly correct, even though in fact it is only approximately correct, and then only for small  $\mathbf{h}$ 's, because there are non-zero higher order terms in the Taylor expansion of  $\nabla f$ .

<sup>2</sup> In the example in the notes, you don't need the caveat about up to a first order approximation, because the higher order terms in the Taylor approximation are all zero. In this example they are not.

<sup>3</sup> The lecture notes tend to change, and sometimes the problem sets don't keep up. If this reference is no longer current, please notify Leo.

- (b) When  $\rho > 0$ , does  $f$  exhibit increasing, constant or decreasing returns to scale? Is your answer true for all  $\rho \in (0, 1]$ . (Again, your answer should be in terms of what happens to the *gradient vector* as you move out along a ray.)

**Ans:** We'll show that  $f$  exhibits decreasing returns to scale whenever  $0 < \rho < 1$ . When  $\rho = 0$ , the function is flat; when  $\rho < 0$ , the function decreases so that the idea of "returns to scale" doesn't mean much. For  $\rho > 0$ , it is sufficient to show that as you increase scale, i.e., move from  $\mathbf{x}$  to  $(1 + \lambda)\mathbf{x}$ , for any  $\lambda > 0$ , the length of the gradient vector shrinks. From the answer to the preceding part, note that for  $d\mathbf{x} = \lambda\mathbf{x}$ , and  $\lambda > 0$ ,  $\nabla f(\mathbf{x} + d\mathbf{x}) \approx (1 + \lambda(\rho - 1)) \nabla f(\mathbf{x})$ . For  $\rho < 1$ ,  $(1 + \lambda(\rho - 1)) < 1$ , so the gradient vector indeed shrinks. When  $\rho = 1$ ,  $(1 + \lambda(\rho - 1)) = 1$ , so the gradient vector remains constant, which is equivalent to constant returns to scale.

- (c) Fix  $\mathbf{x} = (\alpha, \alpha)$ , and consider  $\mathbf{h} = (-0.1, 0.1)$ . Approximate  $\nabla f(\mathbf{x} + \mathbf{h})$ , for (i)  $\rho = 1/2$ ; (ii)  $\rho = -1/2$ ; (iii)  $\rho = -10$ .

**Ans:**

$$\begin{aligned} \nabla f(\mathbf{x} + \mathbf{h}) &\approx \rho\alpha^{\rho-1} \begin{bmatrix} 1 \\ 1 \end{bmatrix} + \begin{bmatrix} \rho(\rho-1)\alpha^{\rho-2} & 0 \\ 0 & \rho(\rho-1)\alpha^{\rho-2} \end{bmatrix} \begin{bmatrix} -0.1 \\ 0.1 \end{bmatrix} \\ &= \begin{cases} \frac{1}{2}\alpha^{-1/2} \left\{ \begin{bmatrix} 1 \\ 1 \end{bmatrix} + 0.1 \begin{bmatrix} \frac{1}{2}\alpha^{-1} \\ -\frac{1}{2}\alpha^{-1} \end{bmatrix} \right\} & \text{if } \rho = 1/2 \\ -\frac{1}{2}\alpha^{-3/2} \left\{ \begin{bmatrix} 1 \\ 1 \end{bmatrix} + 0.1 \begin{bmatrix} -\frac{3}{2}\alpha^{-1} \\ \frac{3}{2}\alpha^{-1} \end{bmatrix} \right\} & \text{if } \rho = -1/2 \\ -10\alpha^{-11} \left\{ \begin{bmatrix} 1 \\ 1 \end{bmatrix} + 0.1 \begin{bmatrix} -11\alpha^{-1} \\ 11\alpha^{-1} \end{bmatrix} \right\} & \text{if } \rho = -10 \end{cases} \end{aligned}$$

- (d) How does the curvature of the level sets of this function change as you move out along a ray through the origin. In particular, discuss the effect of the magnitude of  $\alpha$  on the *rate of change* in the direction of  $\nabla f$  as you add  $\mathbf{h} = (-\beta, \beta)$  to  $\mathbf{x} = (\alpha, \alpha)$ .

**Ans:** Consider the answer to part (c): in each case, the magnitude of the rotation is determined by  $\alpha^{-1}$ . When  $\alpha$  is very large, the rotation is miniscule, when  $\alpha$  is miniscule, the rotation is huge. Thus as you move out along a ray through the origin, the change in curvature generated by a *given* shift  $\mathbf{h} = (-\beta, \beta)$  becomes less and less; that is, level sets become flatter and flatter as you move out along any ray.

(4) Suppose that  $f : \mathbb{R} \rightarrow \mathbb{R}$  is  $(n+1)$  times continuously differentiable.

- (a) Show that a sufficient condition for  $f$  to attain a strict (local) maximum at  $x_0$  is that for some even number  $n$ , the derivatives  $f^{(k)}(x_0)$  are zero for  $k = 1 \dots n - 1$ , and  $f^{(n)}(x_0)$  is negative

**Ans:**

- (1) We are given that  $n$  is even. Hence  $(x - x_0)^n > 0$  for  $x \neq x_0$
- (2) Since  $f^{(n)}(x_0) < 0$  and the first  $n - 1$  terms in the Taylor expansion are all zero, we have that  $T_n(f, x, dx) < 0$ , for all  $dx$ .
- (3) From the Taylor Young theorem, we know that if the  $n$ 'th order Taylor expansion is nonzero for  $dx$  sufficiently small, then for all  $dx$  in some neighborhood  $U$  of 0, the absolute value of the expansion dominates the absolute value of the remainder term.
- (4) Conclude that for all  $dx \in U$ ,  $dx \neq 0$ ,  $f(x + dx) < f(x)$ .

- (b) If  $f^{(k)}(x_0)$  is zero for  $k = 1 \dots n - 1$  and  $f^{(n)}(x_0)$  is non-zero, show that there exists an  $\epsilon$ -neighborhood around  $x_0$  where the absolute value of the  $n^{\text{th}}$ -order Taylor expansion is bigger than the absolute value of the remainder term  $R_n(x)$ .

**Ans:** The  $n$ 'th term in the expansion is  $\frac{f^{(n)}(x_0)(x-x_0)^n}{n!}$ . For some  $\eta \in [x_0, x]$ , the remainder term can be written as  $\frac{f^{(n+1)}(\eta)(x-x_0)^{n+1}}{(n+1)!} = \frac{f^{(n+1)}(\eta)(x-x_0)^n(x-x_0)}{(n+1)n!}$ . We need to show that for  $x$  sufficiently close to  $x_0$ ,  $\left| \frac{f^{(n)}(x_0)(x-x_0)^n}{n!} \right| > \left| \frac{f^{(n+1)}(\eta)(x-x_0)^n(x-x_0)}{(n+1)n!} \right|$ . If  $f^{(n+1)}(\eta) = 0$ , the inequality holds trivially. Assume therefore that  $f^{(n+1)}(\eta) \neq 0$ . Extracting the common term, the required inequality will hold if  $|f^{(n)}(x_0)| > \left| \frac{f^{(n+1)}(\eta)(x-x_0)}{(n+1)} \right|$ . Since  $f_{(n+1)}(\cdot)$  is continuous, and  $[x, x_0]$  is a compact set,  $|f_{(n+1)}(\cdot)|$  attains a maximum on this set. (This is a consequence of Weierstrass theorem.) Let  $\bar{f}_{n+1}$  denote this maximum and pick  $\epsilon < \left| \frac{(n+1)f^{(n)}(x_0)}{\bar{f}_{n+1}} \right|$ . For  $x$  such that  $x - x_0 < \epsilon$ ,

$$\left| \frac{f^{(n+1)}(\eta)(x-x_0)}{(n+1)} \right| < \frac{\epsilon \bar{f}_{n+1}}{(n+1)} < |f^{(n)}(x_0)| \quad \text{as required}$$

- (c) Give a counter example to show that the result in part (a) would be false if the words “for some even  $n$ ” were replaced with “for some  $n > 0$ ”.

**Ans:** Take  $f(x) = -x^3$  and note that  $f_{(1)}(0) = f_{(2)}(0) = 0$  while  $f_{(3)}(0) = -6$ ; This function exhibits the property that for so for the (odd) number  $n = 3$ , the derivatives  $f^{(k)}(x_0)$  are zero for  $k = 1 \dots n - 1$ , and  $f^{(n)}(x_0)$  is negative. In this case, all of the conditions of the theorem are satisfied except that  $n$  is odd. And  $x^3$  isn't maximized at zero.

- (d) Explain carefully, but in as few a words as possible, why the argument in (a) works for even  $n$  but not for odd  $n$ .

**Ans:** If  $n$  is odd, then we cannot conclude, as we did in step (1) of our answer to (a) above, that  $(x - x_0)^{(n)} > 0$  for  $x \neq x_0$ . Therefore we cannot conclude that the sign of the  $n$ 'th term in the series is determined by the sign of  $f_{(n)}$ .

- (e) Show that the  $n^{\text{th}}$ -order Taylor expansion around any point  $x_0$  of a polynomial of degree  $n$  ( i.e. a function of the form  $f(x) = \sum_{k=0}^n \alpha_k x^k$ ) is perfectly accurate, regardless of the magnitude of  $dx$ .

**Ans:** Recall that that for  $g(t) = \alpha t^k$

$$\text{If } n \leq k: g_{(n)}(t) = t * (t - 1) * \dots * (t - n) * \alpha t^{k-n}$$

$$\text{If } n > k: g_{(n)}(t) = 0$$

Hence the  $(n + 1)^{\text{th}}$  derivative of a function  $f(x) = \sum_{k=0}^n \alpha_k x^k$  is equal to zero and the remainder term  $R_n(x) = f_{(n+1)}(\eta) \frac{(x-x_0)^{n+1}}{(n+1)!}$  is zero. And  $R_n(x) = 0$  implies that the expansion is perfectly accurate.

- (f) Show that if  $f$  is an arbitrary polynomial of degree 2, i.e.,  $f(x) = ax^2 + bx + c$ , then for any point  $x_0$ , if you add to  $f(x_0)$  the  $2^{\text{nd}}$ -order Taylor expansion around  $f$ , the expression you get is precisely the original function  $f$ . More precisely, show by writing out the Taylor expansion explicitly, that for arbitrary  $dx$ ,  

$$f(x_0 + dx) = f(x_0) + T_2(f, x_0, dx).$$

**Ans:** Let  $f(x) = ax^2 + bx + c$ , so that the second order Taylor expansion of  $f$  around  $x_0$  is

$$\begin{aligned} & f'(x_0) dx + 0.5f''(x_0) dx^2 \\ = & (2ax_0 + b) dx + 0.5 \times 2a dx^2 \\ = & (2ax_0 + b) dx + a dx^2 \end{aligned}$$

Therefore,

$$\begin{aligned} f(x_0 + dx) &= ax_0^2 + bx_0 + c + (2ax_0 + b) dx + a dx^2 \\ &= a(x_0^2 + 2x_0 dx + dx^2) + b(x_0 + dx) + c \\ &= a(x_0 + dx)^2 + b(x_0 + dx) + c \end{aligned}$$

which is precisely the original function!