## CS 5751: Assignment #1

Due on January 23, 2018

 $Eleazar\ 11:00am$ 

Nam Phung

## Exercise #1

Consider the function  $z = f(x, y) = x^2 + y^2$ . Now, write a pdf file named yourlastname\_hw1\_q2.pdf containing the answers to the following tasks:

(1)

Compute the gradient  $\nabla f(x,y)$ . For this task you need to explain every step of your computation of the gradient. You cannot simply write the vector.

We have  $f(x,y) = x^2 + y^2$ . Since  $\frac{\partial f}{\partial x} = 2x$  and  $\frac{\partial f}{\partial y} = 2y$ , we will have the following:

$$\nabla f(x,y) = \left(\frac{\partial f}{\partial x}, \frac{\partial f}{\partial y}\right) = (2x, 2y)$$

Therefore, the gradient of f(x,y) = (2x,2y).

(2)

Compute the Hessian of  $\nabla f(x,y)$ . For this task you need to explain every step of your computation of the Hessian. You cannot simply write it the matrix.

We have the function  $f(x,y) = x^2 + y^2$ . In part 1, we know that  $\frac{\partial f}{\partial x} = 2x$  and  $\frac{\partial f}{\partial y} = 2y$ . Since  $f: \mathbb{R}^2 \to \mathbb{R}$ , the Hessian matrix of f is a square matrix with a dimension  $2 \times 2$ , that is:

$$\nabla^{2} f = \begin{bmatrix} \frac{\partial^{2} f}{\partial x^{2}} & \frac{\partial^{2} f}{\partial xy} \\ \frac{\partial^{2} f}{\partial y^{2}} & \frac{\partial^{2} f}{\partial y^{2}} \end{bmatrix}$$
$$= \begin{bmatrix} \frac{\partial}{\partial x} 2x & \frac{\partial}{\partial y} 2x \\ \frac{\partial}{\partial x} 2y & \frac{\partial}{\partial y} 2y \end{bmatrix}$$
$$= \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix}$$

(3)

Write the second order Taylor series expansion for f(x,y) around  $x_0 = 0$ . Explain your answer.

We have the Taylor series expansion around  $x_0 = 0$  for  $f: \mathbb{R}^2 \to \mathbb{R}$  as follows:

$$f(x_0 + h) = f(x_0) + \left(\nabla f(x_0)\right)h + \frac{1}{2}\left(h^T\right)\left[\left(\nabla^2 f(x_0)\right)h\right]$$

We know that  $x_0$  and h are vectors with the following values:  $x_0 = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$  and  $h = \begin{bmatrix} h_0 \\ h_1 \end{bmatrix}$ . From section 1 and

2 above, we also know that the gradient  $\nabla f = (2x, 2y)$  and the Hessian of f is  $\nabla^2 f = \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix}$ . Therefore, the Taylor series expansion can be written as follows:

$$f(x_0 + h) = f(x_0) + \left(\nabla f(x_0)\right)h + \left(\frac{1}{2}\right)(h^T)\left[\left(\nabla^2 f(x_0)\right)(h)\right]$$

$$= f(0,0) + \left(\nabla f(0,0)\right)h + \left(\frac{1}{2}\right)(h^T)\left[\left(\nabla^2 f(0,0)\right)(h)\right]$$

$$= 0 + \begin{bmatrix} 0 & 0 \end{bmatrix}\begin{bmatrix} h_0 \\ h_1 \end{bmatrix} + \frac{1}{2}\begin{bmatrix} h_0 & h_1 \end{bmatrix}\left(\begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix}\begin{bmatrix} h_0 \\ h_1 \end{bmatrix}\right)$$

$$= \begin{bmatrix} \frac{h_0}{2} & \frac{h_1}{2} \end{bmatrix}\begin{bmatrix} 2h_0 \\ 2h_1 \end{bmatrix}$$

$$= h_0^2 + h_1^2$$

(4)

Consider Figure 2, which is a plot of f(x, y). What do you observe in this plot? In which direction does the gradient point? Why does this happen?

The plot display a surface of a convex function  $f(x,y) = x^2 + y^2$ . The color of the surface indicates the value z ranging from blue (small value) to red (big value). The arrows on the z-surface is vector field indicating the gradient of f(x,y) at different coordinates (x,y). At the origin (x,y) = (0,0) is the global minimum of this convex function. The gradient of this function is diverging from the origin (0,0). This is because this surface has only 1 minimum at (x,y) = (0,0), as we move further away from this point, the slope of the surface becomes higher, which means the gradient will increase.