

# Numerical Analysis [WI4014TU]

## Assignment 1

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1. Prove for a single-variable function  $f(x)$  that, if  $f$  has a local maximum at  $x_0$  and  $f'(x_0)$  exists, then  $f'(x_0) = 0$  (*Fermat's Theorem*).

**Solution:** According to Fermat's Theorem, if  $f$  has a local maximum at  $c$ , and if  $f'(c)$  exists, then  $f'(c) = 0$ .

To prove:  $f'(c) = 0$

Let  $f(x)$  be a single-variable function having a local maximum at point  $x = c$ . Therefore, in the range where  $c$  exists,  $f(c) \geq f(x)$ .

Let  $h$  be any positive or negative integer with  $h \rightarrow 0$  such that

$$f(c) \geq f(c + h)$$

$$\implies f(c) - f(c + h) \geq 0 \tag{1}$$

Since  $h$  can be either positive or negative. We have two cases at hand —  $h > 0$  or  $h < 0$

For  $h > 0$ : Dividing (1) by  $h$

$$\frac{f(c + h) - f(c)}{h} \leq 0$$

Taking limits from the right-hand side

$$\lim_{h \rightarrow 0^+} \frac{f(c+h) - f(c)}{h} \leq \lim_{h \rightarrow 0^+} 0 = 0$$

Since  $f'(c)$  exists

$$\therefore f'(c) = \lim_{h \rightarrow 0} \frac{f(c+h) - f(c)}{h} = \lim_{h \rightarrow 0^+} \frac{f(c+h) - f(c)}{h} \leq 0$$

Hence,  $f'(c) \leq 0$

Similarly, for  $h < 0$ : Dividing (1) by  $h$  [Note: here  $h < 0$ ]

$$\frac{f(c+h) - f(c)}{h} \geq 0$$

Taking limits from the left-hand side

$$\lim_{h \rightarrow 0^-} \frac{f(c+h) - f(c)}{h} \geq \lim_{h \rightarrow 0^-} 0 = 0$$

Since  $f'(c)$  exists

$$\therefore f'(c) = \lim_{h \rightarrow 0} \frac{f(c+h) - f(c)}{h} = \lim_{h \rightarrow 0^-} \frac{f(c+h) - f(c)}{h} \geq 0$$

Hence,  $f'(c) \geq 0$

Since,  $f'(c) \leq 0$  and  $f'(c) \geq 0$ , Therefore,  $f'(c) = 0$ .

2. Prove that, if a function  $f(x)$  has a local maximum at  $x_0$  and can be expanded in the Taylor series around  $x_0$ , then  $f''(x_0) \leq 0$  (*Converse of the Second Derivative Test*).

**Solution:** Given that  $f(x)$  has a local maximum at  $x_0$ , then  $f'(x_0) = 0$  in a range where  $|x - x_0| < \delta$ .

Since,  $f(x)$  can be expanded in a Taylor series,

Taking the forward Taylor series expansion

$$\Rightarrow f(x) = f(x_0) + f'(x_0)(x-x_0) + \frac{f''(x_0)}{2!}(x-x_0)^2 + \frac{f'''(x_0)}{3!}(x-x_0)^3 + \dots \quad (2)$$

Similarly, taking the backward Taylor series expansion

$$\Rightarrow f(x) = f(x_0) - f'(x_0)(x-x_0) + \frac{f''(x_0)}{2!}(x-x_0)^2 - \frac{f'''(x_0)}{3!}(x-x_0)^3 + \dots \quad (3)$$

Adding (2) and (3), we get

$$2f(x) = 2f(x_0) + 2\frac{f''(x_0)}{2!}(x-x_0)^2 + 2\frac{f'''(x_0)}{2!}(x-x_0)^4 + \dots$$

$$\Rightarrow f(x) - f(x_0) - \frac{f'''(x_0)}{4!}(x-x_0)^4 - \dots = \frac{f''(x_0)}{2!}(x-x_0)^2$$

Since  $x_0$  is the local maximum, then  $f(x) \leq f(x_0)$ . Therefore all the terms in the LHS of the equation will be  $\leq 0$

$$\Rightarrow 0 \geq f(x) - f(x_0) - \frac{f'''(x_0)}{4!}(x-x_0)^4 - \dots = \frac{f''(x_0)}{2!}(x-x_0)^2$$

$$\Rightarrow \frac{f''(x_0)}{2!}(x-x_0)^2 \leq 0$$

$$\therefore f''(x_0) \leq 0$$

3. Use Fermat's Theorem to prove that, if a differentiable function  $u(x, y)$  has a local maximum at  $(x_0, y_0)$ , then  $\nabla u(x_0, y_0) = \mathbf{0}$ . Show that the directional derivative  $D_{\mathbf{v}}u(x_0, y_0) = 0$  for any  $\mathbf{v}$  as well.

**Solution:** Given:  $u(x, y)$  is differentiable and has a local maximum at  $(x_0, y_0)$ , Therefore, there exists a neighbourhood,  $N$ , where  $u(x, y) \leq u(x_0, y_0)$ , for all  $(x, y) \in N \subset \mathbb{R}^2$

Since Fermat's Theorem is defined for single variable functions, we cannot use it directly to prove  $\nabla u(x_0, y_0) = \mathbf{0}$ . Therefore, we need to approach it in a different way.

We know that:

$$\nabla u(x_0, y_0) = \begin{bmatrix} \frac{\partial}{\partial x}u(x_0, y_0) \\ \frac{\partial}{\partial y}u(x_0, y_0) \end{bmatrix} \quad (4)$$

Therefore, if we can prove the individual elements to be equal to 0, then inherently we will prove that

$$\nabla u(x_0, y_0) = \mathbf{0} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

Assume two differentiable, single-variable functions  $f(x)$  and  $g(y)$  defined as follows:

$$f(x) = u(x, y_0)$$

This makes  $f(x)$  a single line function defined over the surface of  $u(x, y)$  where  $y = y_0$

By similar logic, we can define  $g(y)$  as

$$g(y) = u(x_0, y)$$

Now, for  $f(x)$ :

Since,  $x_0$  is the local maximum for  $u(x, y)$  at  $y = y_0$ , and  $f(x)$  is defined for a constant value of  $y = y_0$ , then  $x_0$  is the local maximum for  $f(x)$  as well. Therefore, applying Fermat's Theorem to  $f(x)$  at point  $x_0$

$$\implies \frac{d}{dx}f(x_0) = f'(x_0) = 0$$

$$\implies \frac{\partial}{\partial x} u(x_0, y_0) = 0 \quad (5)$$

Similarly, for  $g(y)$ :

Since,  $y_0$  is the local maximum for  $u(x, y)$  at  $x = x_0$ , and  $g(y)$  is defined for a constant value of  $x = x_0$ , then  $y_0$  is the local maximum for  $g(y)$  as well. Therefore, applying Fermat's Theorem to  $g(y)$  at point  $y_0$

$$\begin{aligned} \implies \frac{d}{dy} g(y_0) &= g'(y_0) = 0 \\ \implies \frac{\partial}{\partial y} u(x_0, y_0) &= 0 \end{aligned} \quad (6)$$

Plugging (5), and (6) in (4), we get

$$\nabla u(x_0, y_0) = \begin{bmatrix} 0 \\ 0 \end{bmatrix} = \mathbf{0}$$

Additionally, from the definition of directional derivative, for any direction vector,  $\mathbf{v}$ , we know that

$$D_{\mathbf{v}} u = \nabla u \cdot \mathbf{v} = |\nabla u| |\mathbf{v}| \cos \theta$$

Since, at point  $(x_0, y_0)$ , we proved that  $\nabla u(x_0, y_0) = \mathbf{0}$ . Therefore  $|\nabla u| = 0$  at  $(x_0, y_0)$ .

$$\therefore D_{\mathbf{v}} u = |\nabla u| |\mathbf{v}| \cos \theta = |0| |\mathbf{v}| \cos \theta = 0$$

4. Let  $H$  be the Hessian matrix and  $\mathbf{v} = \langle p, q \rangle$  a unit vector. Show that  $\mathbf{v}^T H \mathbf{v}$  equals the second directional derivative of  $u$  in the direction of  $\mathbf{v}$ , i.e.  $\mathbf{v}^T H \mathbf{v} = D_{\mathbf{v}}(D_{\mathbf{v}}u)$ .

**Solution:** To prove:

$$\mathbf{v}^T H \mathbf{v} = D_{\mathbf{v}}(D_{\mathbf{v}}u) \quad (7)$$

Let  $\mathbf{v}$  be a column matrix, such that

$$\mathbf{v} = \begin{bmatrix} p \\ q \end{bmatrix}$$

Taking LHS of equation (7):

$$\begin{aligned} \mathbf{v}^T H \mathbf{v} &= \begin{bmatrix} p & q \end{bmatrix} \begin{bmatrix} u_{xx} & u_{xy} \\ u_{xy} & u_{yy} \end{bmatrix} \begin{bmatrix} p \\ q \end{bmatrix} \\ \implies \mathbf{v}^T H \mathbf{v} &= \begin{bmatrix} pu_{xx} + qu_{xy} & pu_{xy} + qu_{yy} \end{bmatrix} \begin{bmatrix} p \\ q \end{bmatrix} \\ \implies \mathbf{v}^T H \mathbf{v} &= \begin{bmatrix} p^2u_{xx} + 2pqu_{xy} + q^2u_{yy} \end{bmatrix} \end{aligned} \quad (8)$$

Now, taking RHS of equation (7) and using the definition of directional derivative:

$$D_{\mathbf{v}}u = \nabla u \cdot \mathbf{v}$$

we get

$$D_{\mathbf{v}}u = p \frac{\partial u}{\partial x} + q \frac{\partial u}{\partial y}$$

This quantity is a scalar quantity. Now,

$$\begin{aligned} D_{\mathbf{v}}(D_{\mathbf{v}}u) &= D_{\mathbf{v}}\left(p \frac{\partial u}{\partial x} + q \frac{\partial u}{\partial y}\right) \\ \implies D_{\mathbf{v}}(D_{\mathbf{v}}u) &= \nabla \left(p \frac{\partial u}{\partial x} + q \frac{\partial u}{\partial y}\right) \cdot \mathbf{v} \\ \implies D_{\mathbf{v}}(D_{\mathbf{v}}u) &= \left\langle \frac{\partial}{\partial x} \left(p \frac{\partial u}{\partial x} + q \frac{\partial u}{\partial y}\right), \frac{\partial}{\partial y} \left(p \frac{\partial u}{\partial x} + q \frac{\partial u}{\partial y}\right) \right\rangle \cdot \langle p, q \rangle \\ \implies D_{\mathbf{v}}(D_{\mathbf{v}}u) &= \langle pu_{xx} + qu_{xy}, pu_{xy} + qu_{yy} \rangle \cdot \langle p, q \rangle \end{aligned}$$

$$\implies D_{\mathbf{v}}(D_{\mathbf{v}}u) = p^2 u_{xx} + 2pq u_{xy} + q^2 u_{yy} \quad (9)$$

Since (8) = (9), Therefore, we have proved that  $\mathbf{v}^T H \mathbf{v} = D_{\mathbf{v}}(D_{\mathbf{v}}u)$ .

5. Use the fact that the intersection of any vertical plane going through the point  $(x_0, y_0)$  and the graph of  $u(x, y)$  is a function of a single variable to finish the proof of Theorem 2.3.1.

**Solution:**

**Theorem 2.3.1:** If a sufficiently smooth  $u(x, y)$  has a local maximum at  $(x_0, y_0)$ , then  $\nabla u(x_0, y_0) = 0$  and the Hessian matrix  $H(x_0, y_0)$  is negative semi-definite.

Consider a vertical plane passing through the point  $(x_0, y_0)$  in any arbitrary direction given by  $\mathbf{v} = \langle p, q \rangle$  where  $p$  and  $q$  are some constant values. Then, the intersection of this plane with the graph of  $u(x, y)$  will yield a single variable function  $\tilde{u}$  which, say, is a function of a variable  $t$  given by

$$\tilde{u}(t) = u(x_0 + tp, y_0 + tq)$$

This new function  $\tilde{u}$  gives the values of the function  $u(x, y)$  along a line in the direction of  $\mathbf{v}$ , and the spatial coordinates are given by  $x = x_0 + tp$  and  $y = y_0 + tq$

Now, the first derivative of  $\tilde{u}(t)$ :

$$\frac{d\tilde{u}}{dt} = \frac{d}{dt}u(x_0 + tp, y_0 + tq)$$

Using chain rule (since  $u$  is not directly dependent on  $t$ )

$$\begin{aligned} \frac{d\tilde{u}}{dt} &= \frac{\partial u}{\partial x} \frac{dx}{dt} + \frac{\partial u}{\partial y} \frac{dy}{dt} \\ \implies \frac{d\tilde{u}}{dt} &= u_x p + u_y q \end{aligned}$$

Since, from equations (4), (5), (6) [see question 3], we know that:

$$\nabla u(x_0, y_0) = \begin{bmatrix} \frac{\partial}{\partial x} u(x_0, y_0) \\ \frac{\partial}{\partial y} u(x_0, y_0) \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} = \mathbf{0}$$

At,  $t = 0$ , we have  $x = x_0$  and  $y = y_0$  which are the local maximum coordinates for  $u$ , and inherently for  $\tilde{u}$  as well (since  $\tilde{u}$  exists along the line at the



intersection of the vertical plane and the graph of  $u$ )

$$\implies \frac{d}{dt}\tilde{u}(0) = u_x p + u_y q = 0$$

Therefore,  $\tilde{u}$  has a critical point at  $t = 0$ .

Now, assuming that a local maximum exists at the location  $t = 0$  for the function  $\tilde{u}$ . Therefore it's second derivative must be  $\leq 0$

$$\implies \frac{d^2\tilde{u}}{dt^2} \leq 0$$

$$\implies \frac{d}{dt}(u_x p + u_y q) \leq 0$$

$$\implies \frac{d}{dt}(u_x p + u_y q) = p \frac{du_x}{dt} + q \frac{du_y}{dt} \leq 0$$

Applying the chain rule to this equation, we get

$$\implies \frac{d}{dt}(u_x p + u_y q) = p \left( \frac{\partial u_x}{\partial x} \frac{dx}{dt} + \frac{\partial u_x}{\partial y} \frac{dy}{dt} \right) + q \left( \frac{\partial u_y}{\partial x} \frac{dx}{dt} + \frac{\partial u_y}{\partial y} \frac{dy}{dt} \right) \leq 0$$

$$\implies \frac{d}{dt}(u_x p + u_y q) = p(u_{xx}p + u_{xy}q) + q(u_{xy}p + u_{yy}q) \leq 0$$

$$\implies \frac{d}{dt}(u_x p + u_y q) = p^2 u_{xx} + 2pq u_{xy} + q^2 u_{yy} \leq 0$$

If we observe, this equation resembles (8) [see question 4], and therefore

$$\implies \frac{d^2\tilde{u}}{dt^2} = \mathbf{v}^T H(x, y) \mathbf{v} \leq 0$$

Since,  $t$  is a critical point and  $(x, y) = (x_0, y_0)$  at  $t = 0$

$$\implies \frac{d^2}{dt^2}\tilde{u}(0) = \mathbf{v}^T H(x_0, y_0) \mathbf{v} \leq 0$$

The result  $\mathbf{v}^T H(x_0, y_0) \mathbf{v} \leq 0$  for any directional vector  $\mathbf{v}$  means that the Hessian matrix satisfies the condition for it to be negative semi-definite.