Math Module 6A Notes

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1 Definitions

1.1 Differentiability of a one variable function

f(x) is **differentiable** at a if and only if the derivative

$$f'(a) = \lim_{h \to 0} \frac{f(a+h) - f(a)}{h}$$
 exists

Furthermore, in this case the graph y = f(x) has a **tangent line** at x = a given by the equation:

$$y = f(a) + f'(a)(x - a)$$

1.2 Tangents of a surface

Consider a function f(x,y). If the below partial derivatives exist, it means that the one variable functions g(x) = f(x,b) and h(y) = f(a,y) have tangent lines at x = a, y = b.

$$f_x(a,b) = \frac{d}{dx}f(x,b)|_{x=a}$$

$$f_y(a,b) = \frac{d}{dx} f(a,y)|_{x=b}$$

1.3 Tangent plane to a surface

A tangent plane should contain both tangent lines. The equation for the tangent plane, if such a plane exists, is:

$$z = f(a,b) + f_x(a,b)(x-a) + f_y(a,b)(y-b)$$

However, there is a tricky difference between the one variable case and several variables.

• In one variable, if the derivative f'(a) exists, then the graph z = f(x) has a tangent line at x = a and it's given by:

$$y = f(a) + f'(a)(x - a)$$

• In two variables, even if $f_x(a,b)$, $f_y(a,b)$ both exist, the graph z = f(x,y) might still not have a tangent plane at (x,y) = (a,b). f might even fail to be continuous there.

1.3.1 Example

Let $f: \mathbb{R}^2 \to \mathbb{R}$ be given by:

$$f(x,y) = \begin{cases} \frac{2xy}{x^2 + y^2} & \text{for } (x,y) \neq (0,0) \\ 0 & \text{for } (x,y) = (0,0) \end{cases}$$

$$f(x,0) = \begin{cases} \frac{2x \cdot 0}{x^2 + 0^2} = 0 & \text{when } x \neq 0 \\ 0 & \text{when } x = 0 \end{cases} = 0 \text{ for all } x \tag{1}$$

$$f_x(0,0) = \frac{d}{dx} f(x,0)|_{x=0}$$

$$= \frac{d}{dx} 0|_{x=0} \quad \therefore (1)$$

$$= 0$$

$$f(0,y) = \begin{cases} \frac{2 \cdot 0 \cdot y}{0^2 + y^2} = 0 & \text{when } y \neq 0 \\ 0 & \text{when } y = 0 \end{cases} = 0 \text{ for all } y$$

$$f_y(0,0) = \frac{d}{dy} f(0,y)|_{y=0}$$

$$= \frac{d}{dy} 0|_{y=0} \quad \therefore (2)$$

$$= 0$$

$$z = f(x,y) = \begin{cases} \frac{2xy}{x^2 + y^2} & \text{for } (x,y) \neq (0,0) \\ 0 & \text{for } (x,y) = (0,0) \end{cases}$$

At (0,0):

$$z = f(0,0) + f_x(0,0)x + f_y(0,0)y = 0$$

1.4 Differentiability of a two variable function

Consider a function f(x,y). Saying f is differentiable at (a,b), means the same as:

$$\frac{|f(x,y) - f(a,b) - f_x(a,b)(x-a) - f_y(a,b)(y-b)|}{||(x,y) - (a,b)||} \to 0, \text{ as } (x,y) \to (a,b)$$

If f(x,y) is differentiable at (a,b), the graph z=f(x,y) has a tangent plane at (x,y)=(a,b), given by:

$$z = f(a,b) + f_x(a,b)(x-a) + f_y(a,b)(x-b)$$

1.4.1 Example

Consider the function:

$$f(x,y) = \begin{cases} \frac{x^3}{x^2 + y^2} & \text{for } (x,y) \neq (0,0) \\ 0 & \text{for } (x,y) = (0,0) \end{cases}$$

Is f differentiable at (0,0)? If so, what is the tangent plane at that point?

$$\lim_{(x,y)\to(0,0)} \frac{|f(x,y)-f(0,0)-f_x(0,0)(x-0)-f_y(0,0)(y-0)|}{||(x,y)-(0,0)||}$$

$$= \lim_{(x,y)\to(0,0)} \frac{|f(x,y)-0-1(x-0)-0(y-0)|}{||(x,y)-(0,0)||}$$

$$= \lim_{(x,y)\to(0,0)} \frac{\left|\frac{x^3}{x^2+y^2}-x\right|}{\sqrt{x^2+y^2}}$$

Approaching (0,0) along y=x gives us:

$$\lim_{(x,y)\to(0,0)} \frac{\left|\frac{x^3}{x^2+y^2} - x\right|}{\sqrt{x^2+y^2}} = \frac{\left|\frac{x^3}{2x^2} - x\right|}{\sqrt{2x^2}}$$
$$= \frac{\frac{1}{2}}{\sqrt{2}|x|}$$
$$= \frac{1}{2\sqrt{2}} \neq 0$$

So, indeed, we do not have:

$$\lim_{(x,y)\to(0,0)}\frac{|f(x,y)-f(0,0)-f_x(0,0)(x-0)-f_y(0,0)(y-0)|}{||(x,y)-(0,0)||}=0$$

I.e. f(x,y) is **not** differentiable at (0,0).

1.5 The gradient vector

Consider $f: A \to \mathbb{R}, A \subset \mathbb{R}^n$, a point $\boldsymbol{a} \in A$, and suppose all partial derivatives $f_{x_k}(\boldsymbol{a}), k = 1, \ldots, n$ exist. The **gradient vector** (grad f)(\boldsymbol{a}) of f at \boldsymbol{a} is the vector:

$$(\operatorname{grad} f)(\boldsymbol{a}) = f_{x_1}(\boldsymbol{a}), f_{x_2}(\boldsymbol{a}), \dots, f_{x_n}(\boldsymbol{a}) \in \mathbb{R}^n$$

1.5.1 Example

With $f: \mathbb{R}^3 \to \mathbb{R}$ given by:

$$f(x, y, z) = xy + yz + zx$$

We have:

$$(\text{grad } f)(x, y, z) = (f_x(x, y, z), f_y(x, y, z), f_z(x, y, z)) = (y + z, x + z, y + x)$$

So, for example:

$$(\text{grad } f)(1,2,3) = (5,4,3)$$

1.5.2 Note

For $f: A \to \mathbb{R}, A \subset \mathbb{R}^n$, such that all partial derivatives exist on A, the gradient (grad f)(\mathbf{x}) = $(f_{x_1}(\mathbf{x}), f_{x_2}(\mathbf{x}), \dots, f_{x_n}(\mathbf{x}))$ is a vector valued function on A, i.e. grad $f: A \to \mathbb{R}^n$. Such functions are also known as **vector fields**.

1.5.3 Rewriting the differentiability condition

With $\boldsymbol{a}=(a,b), \boldsymbol{x}=(x,y),$ we have:

$$f_x(a,b)(x-a) + f_y(a,b)(y-b) = (f_x(a,b), f_y(a,b)) \cdot (x-a,y-b) = (\text{grad } f(a)(x-a))$$

Hence, the differentiability condition:

$$\lim_{(x,y)\to(a,b)} \frac{|f(x,y)-f(a,b)-f_x(a,b)(x-a)-f_y(a,b)(y-b)|}{||(x,y)-(a,b)||} = 0$$

Can be rewritten as:

$$\lim_{x \to a} \frac{|f(x) - f(a) - (\operatorname{grad} f)(a)(x - a)}{||x - a||} = 0$$

This expression also makes sense for a function f of n variables, regardless of n.

1.6 Vector field

For $A \subset \mathbb{R}^n$, a function $\mathbf{F}: A \to \mathbb{R}^n$ is called a **vector field** in \mathbb{R}^n .

1.6.1 Example

Let $f: \mathbb{R}^2 \to \mathbb{R}$ be given by:

$$f(x,y) = x^2 + y^2$$

Then (grad f): $\mathbb{R}^2 \to \mathbb{R}^2$ is given by:

$$F(x,y) = (\text{grad } f)(x,y) = (2x,2y)$$

$$F(0,0) = (0,0)$$

$$F(1,0) = (2,0)$$

$$F(-1,-1) = (-2,-2)$$

1.7 Differentiability in n variables

For $f: A \to \mathbb{R}, A \subset \mathbb{R}^n$, saying that f is **differentiable** at $a \in A$ means the same as:

$$\lim_{\mathbf{x}\to\mathbf{a}}\frac{|f(\mathbf{x})-f(\mathbf{a})-(\operatorname{grad} f)(\mathbf{a})(\mathbf{x}-\mathbf{a})}{||\mathbf{x}-\mathbf{a}||}=0$$

1.7.1 Differentiability implies continuity

Consider $f: A \to \mathbb{R}, A \subset \mathbb{R}^n$, If f is differentiable at $\mathbf{a} \in A$, then f is continuous at \mathbf{a} .

1.7.2 A sufficient condition for differentiability

Consider $f: A \to \mathbb{R}, A \subset \mathbb{R}^n$. If there exists $\delta > 0$ such that all partial derivatives of f are continuous on $\{x \in \mathbb{R}^n : ||x - a| < \delta\}$, then f is differentiable at a.

1.8 Tangents in n variables

For a function $f: A \to \mathbb{R}, A \subset \mathbb{R}^n$, differentiable at $\boldsymbol{a} \in A$, its **tangent space** at $\boldsymbol{x} = \boldsymbol{a}$ is the graph of the function:

$$T(\mathbf{x}) = f(\mathbf{a}) + (\text{grad } f)(\mathbf{a}) \cdot (\mathbf{x} - \mathbf{a})$$

1.9 Chain rule

Consider $g: A \to \mathbb{R}^n, A \subset \mathbb{R}, f: B \to \mathbb{R}, B \subset \mathbb{R}^n$. Suppose g is differentiable at $a \in A$ and suppose f is differentiable at g(a). Then:

$$\frac{d}{dt}f(\boldsymbol{g}(t))|_{t=a} = (\text{grad } f)(\boldsymbol{g}(a)) \cdot \boldsymbol{g}'(a)$$

Let's say n = 2, so with:

$$(x,y) = g(t), \text{ and } z = f(x,y) = f(g(t))$$

The theorem tells us that:

$$\frac{dz}{dt} = \frac{d}{dt} f(\mathbf{g}(t))$$

$$= (\operatorname{grad} f)(\mathbf{g}(t)) \cdot \mathbf{g}'(t)$$

$$= \left(\frac{\partial z}{\partial x}, \frac{\partial z}{\partial y}\right) \cdot \left(\frac{dx}{dt}, \frac{dy}{dt}\right)$$

$$= \frac{\partial z}{\partial x} \frac{dx}{dt} + \frac{\partial z}{\partial y} \frac{dy}{dt}$$

1.9.1 Example

Let:

$$z = f(x, y) = x^2 y, \quad (x, y) = g(t) = (\sin t, t^2)$$

By the chain rule:

$$\frac{dz}{dt} = \frac{\partial z}{\partial x} \frac{dx}{dt} + \frac{\partial z}{\partial y} \frac{dy}{dt}$$
$$= 2xy \cdot \cos t + x^2$$
$$= 2t^2 \sin t \cos t + 2t \sin^2 t$$

1.9.2 Chain rule as a procedure

Let:

$$w = f(x, y, z), \quad (x, y, z) = \boldsymbol{g}(t)$$

By the chain rule:

$$\frac{dw}{dt} = \frac{\partial w}{\partial x} \frac{dx}{dt} + \frac{\partial w}{\partial y} \frac{dy}{dt} + \frac{\partial w}{\partial z} \frac{dz}{dt}$$

We can look at each term as a path in the tree below:



1.10 Laplace equation

Consider a function f(x,y). The Laplace equation is:

$$f_{xx} + f_{yy} = 0$$

Or for a 3 variable function f(x, y, z):

$$f_{xx} + f_{yy} + f_{zz} = 0$$

A function satisfying the Laplace equation is said to be **harmonic**.

1.11 Rate of change

For a real valued function f(x, y):

$$f_x(a,b) = \frac{d}{dx}(x,b)|_{x=a}$$
 measures the rate of change of f as x increases

$$f_y(a,b) = \frac{d}{dx}(a,y)|_{y=b}$$
 measures the rate of change of f as y increases

We can rewrite the above as:

$$f_x(a,b) = \frac{d}{dx}(x,b)|_{x=a}$$

$$= \frac{d}{dt}f(a+t,b)|_{t=0}$$

$$= \frac{d}{dt}f((a,b) + t(1,0))|_{t=0}$$

$$f_y(a,b) = \frac{d}{dy}(a,y)|_{y=b}$$

$$= \frac{d}{dt}f(a,b+t)|_{y=b}$$

$$= \frac{d}{dt}f((a,b) + t(0,1))|_{t=0}$$

1.12 Directional derivative

Consider $f: A \to \mathbb{R}, A \subset \mathbb{R}^n$, a point $a \in A$, and a unit vector $u \in \mathbb{R}^n$. The directional derivative $D_{\boldsymbol{u}}f(\boldsymbol{a})$ of f at \boldsymbol{a} in the direction \boldsymbol{u} , provided the derivative exists, is defined as:

$$D_{\boldsymbol{u}}f(\boldsymbol{a}) = \frac{d}{dt}f(\boldsymbol{a} + t\boldsymbol{u})|_{t=0}$$

1.12.1 Example

For $f(x,y) = x^2y$, find the directional derivative of f at (2,1) in the direction of (1,1).

A unit vector in the direction of (1,1) is:

$$\frac{1}{||(1,1)||}(1,1) = \left(\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}}\right) = \boldsymbol{u}$$

$$D_{u}f(2,1) = \frac{d}{dt}f((2,1) + t\left(\frac{1}{\sqrt{2}}\right), \frac{1}{\sqrt{2}}\right)\Big|_{t=0}$$

$$= \frac{d}{dt}f\left(2 + \frac{t}{\sqrt{2}}, 1 + \frac{t}{\sqrt{2}}\right)\Big|_{t=0}$$

$$= \frac{d}{dt}\left(2 + \frac{t}{\sqrt{2}}\right)^{2}\left(1 + \frac{t}{\sqrt{2}}\right)\Big|_{t=0}$$

$$= \frac{d}{dt}\left(2\left(2 + \frac{t}{\sqrt{2}}\right) \cdot \frac{1}{\sqrt{2}}\left(1 + \frac{t}{\sqrt{2}}\right) + \left(2 + \frac{t}{\sqrt{2}}\right)^{2} \cdot \frac{1}{\sqrt{2}}\right)\Big|_{t=0}$$

$$= 2 \cdot 2 \cdot \frac{1}{\sqrt{2}} \cdot 1 + 2^{2} \cdot \frac{1}{\sqrt{2}}$$

$$= \frac{8}{\sqrt{2}}$$

$$= 4\sqrt{2}$$

1.13 Directional derivatives of differentiable functions

Consider $f: A \to \mathbb{R}, A \subset \mathbb{R}^n, \boldsymbol{a} \in A$, and a unit vector $\boldsymbol{u} \in \mathbb{R}^n$. If f is **differentiable** at \boldsymbol{a} , then:

$$D_{\boldsymbol{u}} f(\boldsymbol{a}) = (\text{grad } f)(\boldsymbol{a}) \cdot \boldsymbol{u}$$

1.13.1 Example

For $f(x,y) = x^2y$, find the directional derivative of f at (2,1) in the direction of (1,1).

Unit vector in the direction of (1,1) is $\mathbf{u} = (\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}})$.

$$f_x(x,y) = 2xy$$
, $f_y(x,y) = x^2$ are both continuous.

Hence, f is differentiable.

$$D_{\boldsymbol{u}}f(2,1) = (\operatorname{grad} f)(2,1) \cdot \boldsymbol{u}$$
$$= (4,4) \cdot \left(\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}}\right)$$
$$= \frac{8}{\sqrt{2}}$$
$$= 4\sqrt{2}$$

1.14 Maximum/minimum and zero directional derivative

Suppose $f(x_1, ..., x_m)$ be differentiable at $\mathbf{a} \in \mathbb{R}^n$, and suppose (grad $f)(\mathbf{a}) \neq \mathbf{0}$. Consider the directional derivative $D_{\mathbf{u}}f(\mathbf{a})$ for different unit vector \mathbf{u} .

- 1. The maximum value of $D_{\boldsymbol{u}}f(\boldsymbol{a})$ is $||\operatorname{grad} f(\boldsymbol{a})||$ and is attained when \boldsymbol{u} points in the direction of grad $f(\boldsymbol{a})$.
- 2. The minimum value of $D_{\boldsymbol{u}}f(\boldsymbol{a})$ is $||-\operatorname{grad} f(\boldsymbol{a})||$ and is attained when \boldsymbol{u} points in the opposite direction of grad $f(\boldsymbol{a})$.
- 3. $D_{\boldsymbol{u}}f(\boldsymbol{a})=0$ if and only if \boldsymbol{u} is orthogonal to grad $f(\boldsymbol{a})$.

1.15 Tangent space

Suppose $f: A \to \mathbb{R}, A \subset \mathbb{R}^n$ is a differentiable at \boldsymbol{a} and with $(\text{grad } f)(\boldsymbol{a}) \neq \boldsymbol{0}$. Let $c = f(\boldsymbol{a})$ and let S be the level set:

$$S = \{ \boldsymbol{x} \in A : f(\boldsymbol{x}) = c \}$$

With the **tangent space** of S at \boldsymbol{a} , we mean the set:

$$T = \{ \boldsymbol{x} \in \mathbb{R}^n : (\text{grad } f)(\boldsymbol{a})(\boldsymbol{x} - \boldsymbol{a}) = 0 \}$$

1.15.1 Example

Let S be the surface given by the equation:

$$x^3 + y^2 - z^2 = 0$$

And let $\boldsymbol{a}=(2,1,3).$ Note that $\boldsymbol{a}\in S.$ What is the tangent space for S at \boldsymbol{a} ?

Let:

$$f(x, y, z) = x^3 + y^2 - z^2$$

S is a level surface of f:

$$S: f(x, y, z) = 0$$

Hence, a normal vector for S at \mathbf{a} is (grad f)(\mathbf{a}):

$$(\text{grad } f)(x, y, z) = (3x^2, 2y, -2z)$$

So $(\operatorname{grad} f)(2,1,3) = (12,2,-6)$. An equation for the tangent space (tangent plane) at \boldsymbol{a} is:

$$12(x-2) + 2(y-1) - 6(z-3) = 0$$
$$6x + y - 3z = 4$$

2 Property of tangents in one variable

In **one variable**, let's look at the size of the **absolute error** compared to |x - a| when we approximate y = f(x) with its tangent line:

$$y = f(a) + f'(a)(x - a)$$

We get:

$$\frac{\text{absolute error}}{|x-a|} = \frac{|f(x) - f(a) - f'(a)(x-a)|}{|x-a|}$$
$$= \left| \frac{f(x) - f(a)}{x-a} - f'(a) \right| \to 0, \text{ as } x \to a$$

2.1 Generalising to two variables

Consider a function f(x, y) of **two** variables. We approximate it near (x, y) = (a, b) with the plane:

$$z = f(a,b) + f_x(a,b)(x-a) + f_y(a,b)(y-b)$$

For this plane to really be a **tangent plane**, we need the same behaviour:

$$\frac{\text{absolute error}}{||(x,y)-(a,b)||} \to 0, \text{ as } (x,y) \to (a,b)$$

I.e.

$$\frac{|f(x,y) - f(a,b) - f_x(a,b)(x-a) - f_y(a,b)(y-b)|}{||(x,y) - (a,b)||} \to 0, \text{ as } (x,y) \to (a,b)$$