

# Calculus 3 Notes

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Date

### **Abstract**

Calculus 3 extends the derivative, limit, and integral to functions of multiple variables. It introduces vectors, vector functions, partial derivatives, the gradient vector, iterated integration, line and surface integrals, directional derivatives and tangent planes, optimization, Lagrange multipliers, and the classical theorems of Green, Gauss, and Stokes. This class was taught by Dr. Kassebaum at AACC.

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# Chapter 1

## Vectors and the Geometry of Space

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### 1.1 Vectors

**Theorem 1.1.1 (Properties of Vectors).** Let  $\vec{a}, \vec{b}, \vec{c} \in \mathbb{R}^n$  be vectors, and let  $c, d \in \mathbb{R}$  be scalars. The following properties are satisfied:

$$\vec{a} + \vec{b} = \vec{b} + \vec{a}$$

$$\vec{a} + (\vec{b} + \vec{c}) = (\vec{a} + \vec{b}) + \vec{c}$$

$$\vec{a} + \vec{0} = \vec{a}$$

$$\vec{a} + (-\vec{a}) = \vec{0}$$

$$c(\vec{a} + \vec{b}) = c\vec{a} + c\vec{b}$$

$$(cd)\vec{a} = c(d\vec{a})$$

$$1\vec{a} = \vec{a}$$

**Definition 1.1.1 (Unit Vector).** A unit vector is any vector  $\vec{v}$  satisfying  $|\vec{v}| = 1$ . In physics, we give  $\hat{v}$  to denote the unit vector in the direction of  $\vec{v}$ .

The vectors  $\langle 1, 0, 0 \rangle$ ,  $\langle 0, 1, 0 \rangle$ , and  $\langle 0, 0, 1 \rangle$  are the **standard basis vectors** of  $\mathbb{R}^3$ . We denote them  $\hat{i}$ ,  $\hat{j}$ , and  $\hat{k}$ , respectively.

For some vector  $\vec{v}$ , the unit vector with the same direction as  $\vec{v}$  is given as

$$\frac{\vec{v}}{|\vec{v}|}$$

This notation is slightly abusive, as multiplication and division are not defined for vectors.

One use of vectors is to denote forces. For example, given some  $\vec{F}_1 = \sqrt{2}\langle -5, 5 \rangle$  and  $\vec{F}_2 = \langle 9\sqrt{3}, 9 \rangle$ , we can find the resultant force vector and its magnitude:

$$\vec{F}_1 + \vec{F}_2 = (9\sqrt{3} - 5\sqrt{2})\hat{i} + (9 + 5\sqrt{2})\hat{j}$$

$$\Rightarrow |\vec{F}_1 + \vec{F}_2| = \sqrt{(9\sqrt{3} - 5\sqrt{2})^2 + (9 + 5\sqrt{2})^2}$$

$$= \sqrt{424 + 90(\sqrt{2} - \sqrt{6})}$$

## 1.2 The Dot Product

**Definition 1.2.1 (The Dot Product).** For  $\vec{a} = \langle a_1, a_2, a_3 \rangle$  and  $\vec{b} = \langle b_1, b_2, b_3 \rangle$ , we define the dot product

$$\vec{a} \cdot \vec{b} = a_1 b_1 + a_2 b_2 + a_3 b_3$$

This is an example of an inner product with the mapping  $\langle \cdot, \cdot \rangle : \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R}$ . The extension from  $\mathbb{R}^3$  to  $\mathbb{R}^n$  is given as

$$\vec{a} \cdot \vec{b} = \sum_{i=1}^n a_i b_i$$

where  $a_i, b_i$  represent components of the vectors  $\vec{a}$  and  $\vec{b}$ .

**Definition 1.2.2 (Orthogonal Vectors).**  $\vec{a}$  and  $\vec{b}$  are orthogonal if and only if  $\langle \vec{a}, \vec{b} \rangle = 0$ , where  $\langle \cdot, \cdot \rangle$  gives an inner product. In calculus 3, this inner product is assumed to always be the dot product, and is generally denoted  $\vec{a} \cdot \vec{b}$ .

**Definition 1.2.3 (Direction Angles).** For some  $\vec{a} = \langle a_1, a_2, a_3 \rangle$ , we give  $\alpha, \beta, \gamma$  as the angles formed with the  $x$ ,  $y$ , and  $z$ -axes. We can define them as

$$\cos \alpha = \frac{a_1}{|\vec{a}|} \quad \cos \beta = \frac{a_2}{|\vec{a}|} \quad \cos \gamma = \frac{a_3}{|\vec{a}|}$$

We also find that

$$\frac{\vec{a}}{|\vec{a}|} = \langle \cos \alpha, \cos \beta, \cos \gamma \rangle$$

As an example, consider  $\vec{a} = \langle 1, 2, 3 \rangle$  and  $\vec{b} = -4\hat{i} + 2\hat{j} - \hat{k}$ . From 1.4.3, we have

$$\vec{a} \cdot \vec{b} = -4 + 4 - 3 = -3$$

$$\vec{b} \cdot \vec{a} = -4 + 4 - 3 = -3$$

$$|\vec{a}|^2 = 1 + 4 + 9 = 14$$

$$|\vec{b}|^2 = 16 + 4 + 1 = 21$$

We see there are some properties of the dot product.

**Theorem 1.2.1 (Properties of the Dot Product).** Let  $\vec{a}, \vec{b}, \vec{c} \in \mathbb{R}^n$  be vectors, and let  $c, d \in \mathbb{R}$  be scalars. The following properties are satisfied:

$$\vec{a} \cdot \vec{a} = |\vec{a}|^2$$

$$\vec{a} \cdot \vec{b} = \vec{b} \cdot \vec{a}$$

$$\vec{a} \cdot (\vec{b} + \vec{c}) = \vec{a} \cdot \vec{b} + \vec{a} \cdot \vec{c}$$

$$(c\vec{a}) \cdot \vec{b} = c(\vec{a} \cdot \vec{b}) = \vec{a} \cdot (c\vec{b})$$

$$\vec{0} \cdot \vec{a} = 0$$

**Theorem 1.2.2 (Alternate Definition of the Dot Product).** Let  $\vec{a}, \vec{b}$  be vectors and let  $\theta$  be the angle between them.

$$\vec{a} \cdot \vec{b} = |\vec{a}||\vec{b}| \cos \theta$$

We can now define new things using the dot product.

**Definition 1.2.4 (Vector Projection).** The projection of a vector  $\vec{v}$  onto another vector  $\vec{u}$  is the component of the vector  $\vec{v}$  in the direction of  $\vec{u}$ . One can imagine two vectors coming from the origin. Forming a right triangle using the top vector as the hypotenuse, the length of the leg that lies along the other vector is the projection of the first vector onto the second vector. The projection of  $\vec{u}$  onto  $\vec{v}$  is

$$\text{proj}_{\vec{v}} \vec{u} = \frac{\vec{v} \cdot \vec{u}}{|\vec{v}|^2} \vec{v}$$

**Definition 1.2.5 (Scalar Projection).** We can define the scalar projection, which is simply the signed magnitude of the vector projection, as seen in 1.2.4. The scalar projection of  $\vec{u}$  onto  $\vec{v}$  is given as

$$\text{comp}_{\vec{v}} \vec{u} = \frac{\vec{v} \cdot \vec{u}}{|\vec{u}|}$$

As an example, we can use the scalar projection to define work. If a force  $\vec{F}$  moves an object from  $P_1$  to  $P_2$ , the displacement vector  $\vec{d} = \overrightarrow{P_1 P_2}$ . The work done by the force over this distance is

$$W = \text{comp}_{\vec{d}} \vec{F} |\vec{d}|$$

which is the product of the component of the force over the distance, times the total distance moved. This simplifies to

$$W = \vec{F} \cdot \vec{d}$$

## 1.3 The Cross Product

We begin by defining the determinant of a  $2 \times 2$  and  $3 \times 3$  matrix.

$$\begin{vmatrix} a & b \\ c & d \end{vmatrix} \equiv ad - bc$$

$$\begin{vmatrix} a_1 & a_2 & a_3 \\ b_1 & b_2 & b_3 \\ c_1 & c_2 & c_3 \end{vmatrix} = a_1 \begin{vmatrix} b_2 & b_3 \\ c_2 & c_3 \end{vmatrix} - a_2 \begin{vmatrix} b_1 & b_3 \\ c_1 & c_3 \end{vmatrix} + a_3 \begin{vmatrix} b_1 & b_2 \\ c_1 & c_2 \end{vmatrix}$$

**Definition 1.3.1 (The Cross Product).** The cross product is a vector operation defined with the

mapping  $\mathbb{R}^3 \times \mathbb{R}^3 \rightarrow \mathbb{R}^3$ , where for some  $\vec{a} = \langle a_1, a_2, a_3 \rangle$  and  $\vec{b} = \langle b_1, b_2, b_3 \rangle$ , we give

$$\vec{a} \times \vec{b} = \langle a_2b_3 - a_3b_2, a_3b_1 - a_1b_3, a_1b_2 - a_2b_1 \rangle$$

If we recall the way to evaluate a  $3 \times 3$  determinant from above, we have the useful memorization method:

$$\vec{a} \times \vec{b} = \begin{vmatrix} \hat{\mathbf{i}} & \hat{\mathbf{j}} & \hat{\mathbf{k}} \\ a_1 & a_2 & a_3 \\ b_1 & b_2 & b_3 \end{vmatrix}$$

This isn't really a valid determinant to evaluate, but if we treat it as such then it's easier to remember.

The cross product will always give a vector orthogonal to the two starting vectors. We also find that  $\vec{a} \times \vec{a} \equiv \vec{0}$ . If for some  $\vec{a} \times \vec{b}$ , the angle between the two vectors  $\theta \in [0, \pi]$  goes counter-clockwise, then the resulting vector will point in the positive direction. If the angle goes clockwise, it will point in the negative direction (the negative of its magnitude).

**Theorem 1.3.1** (Sinusoidal Definition of the Cross Product).

$$|\vec{a} \times \vec{b}| = |\vec{a}||\vec{b}| \sin \vartheta$$

**Theorem 1.3.2** (Properties of the Cross Product). Take  $|\varphi\rangle$ ,  $|\psi\rangle$ , and  $|\vartheta\rangle$  as vectors in  $\mathbb{R}^3$ , because  $\vec{a}$  and  $\vec{b}$  have different heights and it's annoying. Let  $\alpha \in \mathbb{R}$ . The following statements will hold:

$$\begin{aligned} |\psi\rangle \times |\varphi\rangle &= -|\varphi\rangle \times |\psi\rangle \\ (\alpha|\psi\rangle) \times |\varphi\rangle &= |\psi\rangle \times (\alpha|\varphi\rangle) = \alpha(|\psi\rangle \times |\varphi\rangle) \\ |\psi\rangle \times (|\varphi\rangle + |\vartheta\rangle) &= |\psi\rangle \times |\varphi\rangle + |\psi\rangle \times |\vartheta\rangle \\ |\psi\rangle \cdot (|\varphi\rangle \times |\vartheta\rangle) &= (|\psi\rangle \times |\varphi\rangle) \cdot |\vartheta\rangle \\ |\psi\rangle \times (|\varphi\rangle \times |\vartheta\rangle) &= (|\psi\rangle \cdot |\vartheta\rangle)|\varphi\rangle - (|\psi\rangle \cdot |\varphi\rangle)|\vartheta\rangle \end{aligned}$$

**Remark.** using this notation for calculus was a mistake.

## 1.4 Equations of Lines and Planes

**Theorem 1.4.1** (Vector Parametrization of Lines). Given a point  $P$  and a direction vector  $\vec{v}$ , we define a line as the set of terminal points of position vectors  $\vec{r}$ . If  $P = (p_1, \dots, p_n)$ , we give  $\vec{r}_0 = \langle p_1, \dots, p_n \rangle$ . We give the vector equation of a line as

$$\vec{r} = \vec{r}_0 + t\vec{v}, t \in \mathbb{R}$$

For example, take the point  $P = (2, 3)$  and  $\vec{v} = \langle 4, 1 \rangle$ . We have  $\vec{r} = \langle 2, 3 \rangle + \langle 4t, t \rangle$ . Since we're defining a line as the set of terminal points of  $\vec{r}$ , we can write  $\langle x, y \rangle = \langle 2 + 4t, 3 + t \rangle$ . It follows from the definition of matrix addition that we have  $x = 2 + 4t$  and  $y = 3 + t$ . These are the parametric equations of this line. We solve for  $t$  and have  $t = \frac{x-2}{4} = y - 3$ . These are the symmetric equations

of this line.

Sidenote: if  $f : \langle x_1, \dots, x_n \rangle \mapsto (x_1, \dots, x_n)$ , then we can give our line in set-builder notation as

$$\{f(\vec{r}) \mid \vec{r} = \vec{r}_0 + t\vec{v}, t \in \mathbb{R}\}$$

We will be using  $P \sim \vec{r}$  and  $\vec{r} \sim P$  to notate the position vector  $\vec{r}$  for a point  $P$ .

**Definition 1.4.1 (Direction Numbers).** If we take  $\vec{v} = \langle \alpha, \beta, \gamma \rangle$  in the above example, then the direction numbers are  $\alpha, \beta, \gamma$  since these are the denominators of our symmetric equations.

**Theorem 1.4.2 (Defining Line Segments with Position Vectors).** The line segment between points  $P_0$  and  $P_1$  with corresponding position vectors  $P_0 \sim \vec{r}_0$  and  $P_1 \sim \vec{r}_1$  is given by

$$\vec{r}(t) = (1 - t)\vec{r}_0 + t\vec{r}_1, t \in [0, 1] \subset \mathbb{R}$$

We can verify the above theorem trivially. We have the parametric variable  $t$  spanning the interval  $[0, 1]$ , so we must have  $t = 0 \Rightarrow \vec{r}(0) = \vec{r}_0$  and  $t = 1 \Rightarrow \vec{r}(1) = \vec{r}_1$ . Consider the following:

$$\vec{r}(t) = (1 - t)\vec{r}_0 + t\vec{r}_1$$

$$\vec{r}(0) = (1 - 0)\vec{r}_0 + 0\vec{r}_1$$

$$\Rightarrow \vec{r}(0) = \vec{r}_0$$

$$\vec{r}(1) = (1 - 1)\vec{r}_0 + 1\vec{r}_1$$

$$\Rightarrow \vec{r}(1) = \vec{r}_1$$

**Definition 1.4.2 (Planes).** A plane in  $\mathbb{R}^3$  is determined by some point  $P_0 = (x_0, y_0, z_0)$  and a normal vector  $\vec{n}$  which is orthogonal to the plane.

For example, consider the point  $P_0 = (0, 2, 3)$  and vector  $\vec{n} = \langle 2, 0, 0 \rangle$ . We find  $\vec{n} \sim \hat{i}$ , and  $P_0 \in \mathbb{R}_y \times \mathbb{R}_z$ , thus this vector is orthogonal to the  $yz$  plane, which the point happens to be in. Thus this vector and point describe the  $yz$ -plane.

Now consider the point  $P_0 = (0, 2, 3)$  but the vector  $\vec{n} = \langle 2, 0, 1 \rangle$ . If we define the plane described herein as the set of points  $\psi$ , then we can define  $P = (x, y, z) \neq P_0 \wedge P \in \psi$ . Let  $\vec{r}_0 \sim P_0$  be the position vector of point  $P_0$ , and let  $\vec{r} \sim P$  be the position vector of point  $P$ . This means

$$\vec{r}_0 = \langle 0, 2, 3 \rangle$$

$$\vec{r} = \langle x, y, z \rangle$$

We define a new vector  $\vec{r} - \vec{r}_0 = \overrightarrow{PP_0}$  which is parallel to the plane. Thus,

$$\vec{n} \cdot (\vec{r} - \vec{r}_0) = 0$$

is the vector equation of the plane. We also find

$$\Rightarrow \langle 2, 0, 1 \rangle \cdot \langle x - 0, y - 2, z - 3 \rangle = 0$$

$$\Rightarrow 2x + z - 3 = 0$$



This is the scalar equation of the plane. Notice this is a linear equation in  $x, y, z$ . In fact, any linear equation  $ax + by + cz + d = 0$  has a plane with normal vector  $\vec{n} = \langle a, b, c \rangle$ .

Another thing we can do is find a vector normal to the plane given points. For example, consider  $P = (1, 3, 0)$ ,  $Q = (2, -1, 5)$ , and  $R = (4, 0, 2)$ . We can give the vectors

$$\overrightarrow{PR} = \langle -3, 3, 2 \rangle, \quad \overrightarrow{QP} = \langle 1, -4, 5 \rangle$$

We compute  $\overrightarrow{PR} \times \overrightarrow{QP}$  to find a vector normal to the plane, since we know  $\overrightarrow{QP}$  and  $\overrightarrow{PR}$  are both in the plane.

$$\begin{aligned} \overrightarrow{PR} \times \overrightarrow{QP} &= \begin{vmatrix} \hat{i} & \hat{j} & \hat{k} \\ -3 & 3 & -2 \\ 1 & -4 & 5 \end{vmatrix} \\ &= \begin{vmatrix} 3 & -2 \\ -4 & 5 \end{vmatrix} \hat{i} - \begin{vmatrix} -3 & -2 \\ 1 & 5 \end{vmatrix} \hat{j} + \begin{vmatrix} -3 & 3 \\ 1 & -4 \end{vmatrix} \hat{k} \\ &= (15 - 8)\hat{i} - (-15 + 2)\hat{j} + (12 - 3)\hat{k} \end{aligned}$$

**Definition 1.4.3 (Angle Between Planes).** For planes  $\psi$  and  $\phi$ , the angle between the planes  $\theta \in [0, \pi]$  is given by the angle  $\theta$  between the normal vectors  $\vec{n}_\psi$  and  $\vec{n}_\phi$ .

**Definition 1.4.4 (Parallel and Orthogonal Planes).** For planes  $\psi$  and  $\phi$ , they are parallel if and only if their normal vectors  $\vec{n}_\psi$  and  $\vec{n}_\phi$  are parallel. Furthermore,  $\phi \perp \psi$  if and only if  $\vec{n}_\phi \cdot \vec{n}_\psi = 0$ .

## 1.5 Cylinders and Quadric Surfaces

**Definition 1.5.1 (Traces).** A trace of a surface is the curve of its intersection with a coordinate plane.

**Definition 1.5.2 (Cylinder).** A surface consisting of all lines parallel to it that pass through a given plane curve.

**Definition 1.5.3 (Quadric Surface).** A surface defined by a polynomial equation of degree 2, that is an equation of the form

$$Ax^2 + By^2 + Cz^2 + Dxy + Eyz + Fxz + Gx + Hy + Iz + J = 0, \quad A, B, C, D, E, F, G, H, I, J \in \mathbb{R}$$

Basic quadric surfaces can be defined of the forms

- $z = Ax^2 + By^2$  describes a paraboloid
- $z^2 = Ax^2 + By^2$  describes a double cone
- $\frac{x^2}{A^2} + \frac{y^2}{B^2} - \frac{z^2}{C^2} = 1$  describes a hyperboloid of one sheet

- 
- $-\frac{x^2}{A^2} - \frac{y^2}{B^2} + \frac{z^2}{C^2} = 1$  describes a hyperboloid of two sheets
  - $\frac{x^2}{A^2} + \frac{y^2}{B^2} + \frac{z^2}{C^2} = 1$  describes an ellipsoid

# Chapter 2

## Vector Functions

### 2.1 Vector Functions and Space Curves

**Definition 2.1.1 (Vector-Valued Function).** A vector-valued function  $\vec{r}: \mathbb{R} \rightarrow \mathbb{R}^n$  has a domain of real numbers and a range of vectors.

**Definition 2.1.2 (Component Functions).** Component functions  $f, g, h$  of a vector function  $\vec{r}$  are given as

$$\vec{r}(t) = f(t)\hat{\mathbf{i}} + g(t)\hat{\mathbf{j}} + h(t)\hat{\mathbf{k}}$$

**Definition 2.1.3 (Space Curve).** A space curve is the set  $C$  of points  $(x, y, z)$  where

$$C = \{(x, y, z) \mid t \in I \subseteq \mathbb{R}, x = f(t), y = g(t), z = h(t)\}$$

Let  $\vec{r}(t) = \langle 1 + 3t, 2 - 4t, 7 + t \rangle$ . It follows that

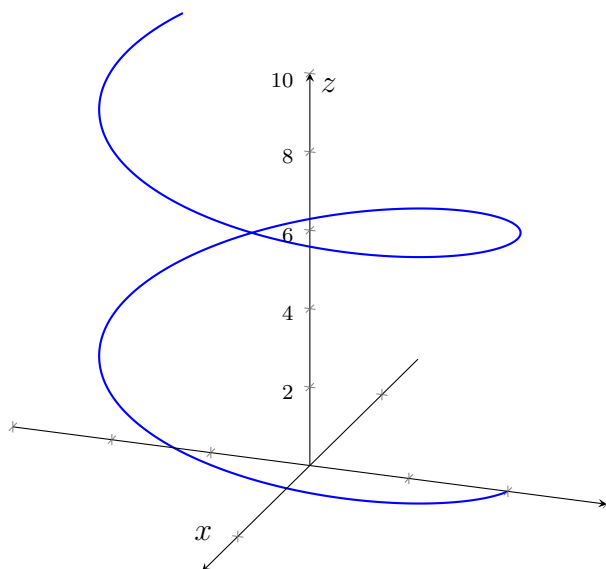
$$\langle x, y, z \rangle = \langle 1 + 3t, 2 - 4t, 7 + t \rangle$$

$$\Rightarrow x = 1 + 3t, \quad y = 2 - 4t, \quad z = 7 + t$$

These are the **parametric equations** of this vector function. We can then also show that

$$t = \frac{x - 1}{3} = \frac{y - 2}{-4} = z - 7$$

These are the **symmetric equations** of this vector function.



$$\vec{r}(t) = \sin(t)\hat{\mathbf{i}} + \cos(t)\hat{\mathbf{j}} + t\hat{\mathbf{k}}, \quad t \in \mathbb{R}^+ \cup \{0\}$$

**Definition 2.1.4 (Limit of a Vector Function).** If  $\vec{r}(t) = \langle f(t), g(t), h(t) \rangle$ , then we will have

$$\lim_{t \rightarrow a} \vec{r}(t) = \lim_{t \rightarrow a} \langle f(t), g(t), h(t) \rangle = \langle \lim_{t \rightarrow a} f(t), \lim_{t \rightarrow a} g(t), \lim_{t \rightarrow a} h(t) \rangle$$

provided the limits of the component functions exist.

## 2.2 Derivatives and Integrals of Vector Functions

**Definition 2.2.1 (Derivative of a Vector Function).** For a vector function  $\vec{r}(t) = \langle f(t), g(t), h(t) \rangle, t \in \mathbb{R}$ , we define

$$\vec{r}' = \lim_{h \rightarrow 0} \frac{\vec{r}(t+h) - \vec{r}(t)}{h}$$

provided this limit exists.

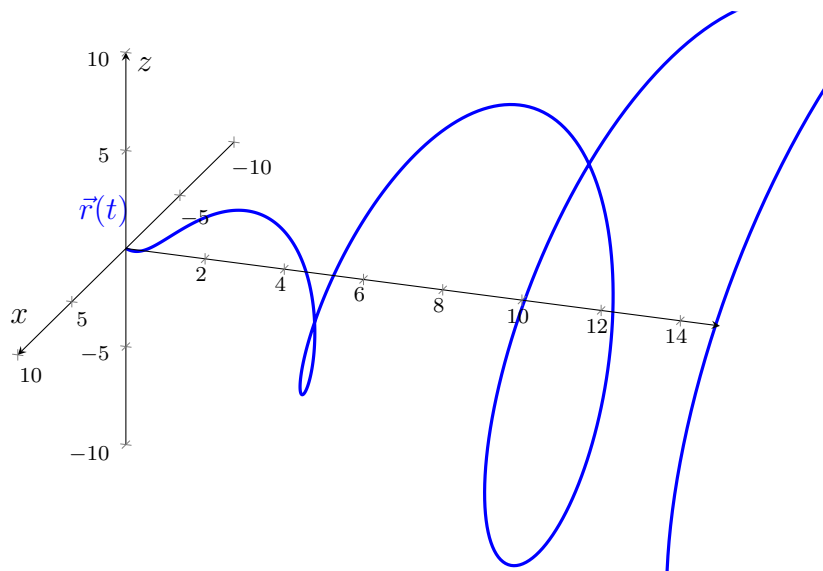
From our earlier definition of the limit of a vector function, we have

$$\lim_{h \rightarrow 0} \frac{\vec{r}(t+h) - \vec{r}(t)}{h} = \left\langle \lim_{h \rightarrow 0} \frac{f(t+h) - f(t)}{h}, \lim_{h \rightarrow 0} \frac{g(t+h) - g(t)}{h}, \lim_{h \rightarrow 0} \frac{h(t+h) - h(t)}{h} \right\rangle$$

We call this vector the **tangent vector**.

For example, consider the following space curve:

$$\vec{r}(t) = \langle t \cos(t), t, t \sin(t) \rangle$$



We give

$$\vec{r}'(t) = \langle \cos(t) - t \sin(t), 1, \sin(t) + t \cos(t) \rangle$$

**Definition 2.2.2 (Tangent Line).** The tangent line to the curve traced by  $\vec{r}(t)$  at point  $P = (x_0, y_0, z_0)$  is the line parallel to the tangent vector and through point  $P$ . We give the following definition:

$$\frac{x - x_0}{f'(t)} = \frac{y - y_0}{g'(t)} = \frac{z - z_0}{h'(t)}$$

**Definition 2.2.3 (Unit Tangent Vector).** We often use the **Unit Tangent Vector**,  $\vec{T}(t)$ , where

$$\vec{T}(t) = \frac{\vec{r}'(t)}{|\vec{r}'(t)|}$$

Example: find the unit tangent vector at  $t = 0$  for

$$\vec{r}(t) = t^3 e^t \hat{\mathbf{i}} - 4t^3 \hat{\mathbf{j}} + \sin(3t) \hat{\mathbf{k}}$$

$$\vec{r}'(t) = (3t^2 e^{t+t^3} e^t) \hat{\mathbf{i}} - 12t^2 \hat{\mathbf{j}} + 3 \cos(3t) \hat{\mathbf{k}}$$

$$\implies \vec{r}'(0) = 3 \hat{\mathbf{k}}$$

If we changed the  $\hat{\mathbf{k}}$  component to  $\cos$  instead of  $\sin$ , we would have  $\vec{r}'(0) = \vec{0}$ , meaning the derivative is undefined for  $t = 0$ .

**Definition 2.2.4 (Differentiation Rules for Vector-Valued Functions).** Let  $\vec{v}(t), \vec{u}(t)$  be differentiable vector functions,  $c \in \mathbb{R}$ , and  $f$  is a real-valued function.

- $\frac{d}{dt}(\vec{u}(t) + \vec{v}(t)) = \frac{d}{dt}\vec{u}(t) + \frac{d}{dt}\vec{v}(t)$
- $\frac{d}{dt}(c\vec{u}(t)) = c\vec{u}'(t)$
- $\frac{d}{dt}(f(t)\vec{u}(t)) = f(t)\vec{u}'(t) + f'(t)\vec{u}(t)$
- $\frac{d}{dt}(\vec{v}(t) \cdot \vec{u}(t)) = \vec{u}'(t) \cdot \vec{v}(t) + \vec{v}'(t) \cdot \vec{u}(t)$
- $\frac{d}{dt}(\vec{u}(t) \times \vec{v}(t)) = \vec{u}'(t) \times \vec{v}(t) + \vec{u}(t) \times \vec{v}'(t)$
- $\frac{d}{dt}\vec{u}(f(t)) = f'(t)\vec{u}'(f(t))$

For all of these, order of operation will only matter for the cross product.

**Definition 2.2.5 (Integrals of Vector Functions).** For  $t \in I = [\alpha, \beta] \subseteq \mathbb{R}$  we give

$$\int_I \vec{r}(t) dt = \left\langle \int_{\alpha}^{\beta} f(t) dt, \int_{\alpha}^{\beta} g(t) dt, \int_{\alpha}^{\beta} h(t) dt \right\rangle + \vec{C}$$

## 2.3 Arc Length and Curvature

**Definition 2.3.1 (Arc Length Function  $s$  in 3 Dimensions).** For  $\vec{r}(t) = \langle f(t), g(t), h(t) \rangle, t \in [a, b] \subseteq \mathbb{R}$ , we have

$$s(t) = \int_a^t |\vec{r}'(u)| du = \int_a^t \sqrt{\left(\frac{dx}{du}\right)^2 + \left(\frac{dy}{du}\right)^2 + \left(\frac{dz}{du}\right)^2} du$$

**Definition 2.3.2 (Smooth Parametrization of a Curve).**  $\vec{r}(t)$  is smooth for an interval  $t \in I$  if  $\vec{r}'$  is continuous and  $\vec{r}'(t) \neq \vec{0} \forall t \in I$ .

**Definition 2.3.3 (Curvature (The intuitive and useless definition)).**

$$\kappa = \left| \frac{d\vec{T}}{ds} \right|$$

Suppose we have  $|\vec{r}'(t)| = c \in \mathbb{R} \forall t$ . We then have

$$\begin{aligned} \vec{r}(t) \cdot \vec{r}(t) &= |\vec{r}(t)|^2 = c^2 \\ \implies \frac{d\vec{r}}{dt} &= 0 = 2(\vec{r}'(t) \cdot \vec{r}(t)) \\ \left| \vec{T}(t) \right| &= 1 \forall t \\ \therefore \vec{T}(t) &\perp \vec{T}'(t) \end{aligned}$$

**Definition 2.3.4** (Principal Unit Normal Vector (also Unit Normal)). The unit normal  $\vec{N}(t)$  is given as

$$\vec{N}(t) = \frac{\vec{T}'(t)}{|\vec{T}'(t)|}$$

**Definition 2.3.5** (Binormal Vector). The binormal vector  $\vec{B}(t)$  is given as

$$\vec{B}(t) = \vec{T}(t) \times \vec{N}(t)$$

**Definition 2.3.6** (Torsion of a Curve). Torsion generally indicates how tight a curve bends. We give it as  $\tau$ .

$$\tau = -\frac{d\vec{B}}{ds} \cdot \vec{N}(t) = -\frac{\vec{B}(t) \cdot \vec{N}(t)}{|\vec{r}'(t)|}$$

Recall that for some  $(x, y) = (f(t), g(t))$ , arclength  $s$  is given as

$$s = \int_a^b \sqrt{(f'(t))^2 + (g'(t))^2} dt$$

In three dimensions with a curve parametrized by a vector function where  $t \in [a, b]$ , we have

$$s(t) = \int_a^t |\vec{r}'(u)| dt$$

By the fundamental theorem of calculus, we also conclude that

$$\frac{ds}{dt} = |\vec{r}'(t)|$$

From the definition of the magnitude of a vector, we have

$$s(t) = \int_a^t \sqrt{(f'(u))^2 + (g'(u))^2 + (h'(u))^2} dt$$

Since  $s$  is a function of  $t$ , we can often reparametrize  $\vec{r}$  as a function of its length, which can be exceedingly useful, but rather tough. To do this, we compute the integral definition of arc length, solve for  $t$  in terms of  $s$ , and substitute our solution in for  $t$  in the original  $\vec{r}$ .

An inverse function does not always exist though, and we need the implicit function theorem to tell us if it does. So this is not always possible to do. However, it's possible to *assume* we can reparametrize something that we can't. Recall curvature  $\kappa = \left| \frac{d\vec{T}}{ds} \right|$ . From our knowledge of the derivatives of parametric functions, and the fact that  $\vec{T}$  is a function of  $s$ , even though it may not always be possible to write it as one, we give

$$\frac{d\vec{T}}{dt} = \frac{d\vec{T}}{ds} \frac{ds}{dt}$$

An easier way of thinking about this is as the chain rule:

$$\frac{d}{dt} \vec{T}(s(t)) = \vec{T}'(s(t)) s'(t)$$

We can now solve for  $\frac{d\vec{T}}{ds}$ :

$$\frac{d\vec{T}}{dt} = \frac{\vec{T}}{ds} \frac{ds}{dt} \implies \frac{d\vec{T}}{\frac{ds}{dt}} = \frac{d\vec{T}}{ds}$$

Recall  $\kappa = \left| \frac{d\vec{T}}{ds} \right|$  and  $\frac{ds}{dt} = |\vec{r}'(t)|$ . We have

$$\kappa(t) = \left| \frac{\frac{d\vec{T}}{dt}}{|\vec{r}'(t)|} \right|$$

This re-definition of curvature is sometimes useful.

**Definition 2.3.7 (Curvature (The Useful One)).** The useful definition of curvature:

$$\kappa(t) = \frac{|\vec{r}'(t) \times \vec{r}''(t)|}{|\vec{r}'(t)|^3}$$

For example, consider the curvature of the curve  $C$  parameterized by  $\vec{r}(t) = \langle t, t^2, t^3 \rangle$ .

$$\langle t, t^2, t^3 \rangle = \langle 2, 4, 8 \rangle \iff t = 2\vec{r}'(t) = \langle 1, 2t, 3t^2 \rangle$$

$$\vec{r}'(2) = \langle 1, 4, 12 \rangle$$

$$\vec{r}''(t) = \langle 0, 2, 6t \rangle$$

$$\vec{r}''(2) = \langle 0, 2, 12 \rangle$$

$$\begin{vmatrix} \hat{\mathbf{i}} & \hat{\mathbf{j}} & \hat{\mathbf{k}} \\ 1 & 4 & 12 \\ 0 & 2 & 12 \end{vmatrix} = \begin{vmatrix} 4 & 12 \\ 2 & 12 \end{vmatrix} \hat{\mathbf{i}} - \begin{vmatrix} 1 & 12 \\ 0 & 12 \end{vmatrix} \hat{\mathbf{j}} + \begin{vmatrix} 1 & 4 \\ 0 & 2 \end{vmatrix} \hat{\mathbf{k}} = 24\hat{\mathbf{i}} - 12\hat{\mathbf{j}} + 2\hat{\mathbf{k}}$$

$$\implies \frac{|\vec{r}'(t) \times \vec{r}''(t)|}{|\vec{r}'(t)|^3} = \frac{\sqrt{24^2 + 12^2 + 2^2}}{(1 + 16 + 144)^{\frac{3}{2}}} \approx 0.0131713554$$

$$\therefore \kappa(2) \approx 0.0131713554$$

Consider a space curve, in this case parametrized by  $3\sin(t)\hat{\mathbf{i}} + 3\cos(t)\hat{\mathbf{j}} + 4t\hat{\mathbf{k}}$ . Consider the point on the curve  $t_0 = \frac{\pi}{2}$ .

$$\vec{r}(t_0) = 3\hat{\mathbf{i}} + 2\pi\hat{\mathbf{k}}$$

$$\vec{r}'(t) = 3\cos(t)\hat{\mathbf{i}} - 3\sin(t)\hat{\mathbf{j}} + \hat{\mathbf{k}}$$

$$\vec{r}'(t_0) = -3\hat{\mathbf{j}} + 4\hat{\mathbf{k}}$$

$$\vec{T}(t) = \frac{3\cos(t)\hat{\mathbf{i}} - 3\sin(t)\hat{\mathbf{j}} + 4\hat{\mathbf{k}}}{\sqrt{9(\cos^2(t) + \sin^2(t)) + 16}}$$

$$\vec{T}(t) = \frac{1}{5} (3\cos(t)\hat{\mathbf{i}} - 3\sin(t)\hat{\mathbf{j}} + 4\hat{\mathbf{k}})$$

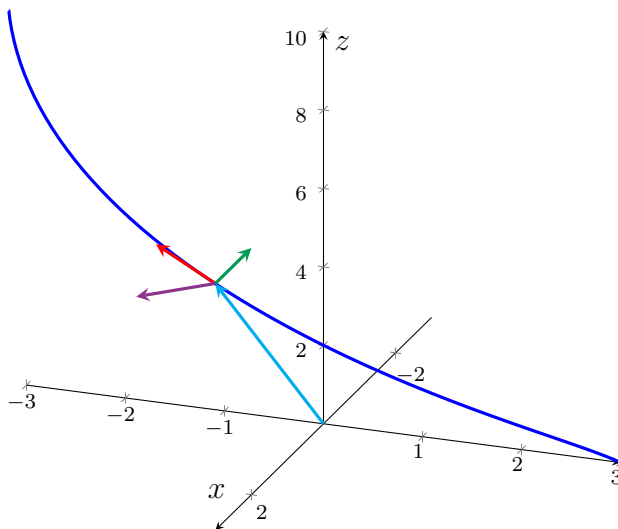
$$\vec{T}(t_0) = \frac{1}{5} \langle 0, -3, 4 \rangle$$

$$\vec{T}'(t) = \frac{1}{5} (-3\sin(t)\hat{\mathbf{i}} - 3\cos(t)\hat{\mathbf{j}})$$

$$\vec{T}'(t_0) = \left\langle -\frac{3}{5}, 0, 0 \right\rangle$$



$$\begin{aligned}
\vec{N}(t_0) &= -\hat{\mathbf{i}} \\
\vec{B}(t_0) &\sim \begin{vmatrix} \hat{\mathbf{i}} & \hat{\mathbf{j}} & \hat{\mathbf{k}} \\ 0 & -\frac{3}{5} & \frac{4}{5} \\ -1 & 0 & 0 \end{vmatrix} \\
&= -\frac{4}{5}\hat{\mathbf{j}} - \frac{3}{5}\hat{\mathbf{k}}
\end{aligned}$$



We can create a basis for  $\mathbb{R}^3$  due to the fact that  $\vec{T} \perp \vec{N} \perp \vec{B}$ . Computing all these vectors for the general case can be **hard**, so we often don't.

**Definition 2.3.8 (Osculating Plane).** The tangent plane to a curve, parallel to the unit tangent and orthogonal to the binormal vector.

**Definition 2.3.9 (Normal Plane).** The plane orthogonal to the curve at all points, and thus orthogonal to the unit tangent.

## 2.4 Velocity and Acceleration

Let  $\vec{r}(t)$  be a position function for a particle with respect to the parameter time. Likewise, let  $\vec{v}$  and  $\vec{a}$  define velocity and acceleration. Let  $v$  define speed.

$$\begin{aligned}
\vec{T}(t) &= \frac{\vec{r}'(t)}{|\vec{r}'(t)|} = \frac{\vec{v}}{|\vec{v}|} = \frac{\vec{v}}{v} \\
&\implies \vec{v} = v\vec{T} \\
&\implies \vec{a} = v'\vec{T} + v\vec{T}'
\end{aligned}$$

---


$$\text{Recall } \kappa(t) = \frac{|\vec{T}'|}{|\vec{r}'|} = \frac{|\vec{T}'|}{v}$$

$$\implies |\vec{T}'| = v\kappa$$

$$\implies \vec{N} = \frac{\vec{T}'}{v\kappa} \implies \vec{T}' = \vec{N}v\kappa$$

$$\implies v\vec{T}' + v^2\kappa\vec{N}$$

We now see that acceleration always occurs in the plane parallel to the unit tangent and unit normal vectors, and parallel to the binormal vector. Thus, acceleration only happens in the osculating plane. This process is known as vector decomposition.

**Definition 2.4.1 (Components of Acceleration).** We give  $a_T = v'$  and  $a_N = v^2\kappa$ , thus

$$\vec{a} = a_T\vec{T} + a_N\vec{N}$$

Also keep in mind that when integrating a vector-valued function representing acceleration or velocity, we can't simply add  $v_0$  or  $r_0$ , we need to calculate the value for a constant such that it will come out to that at  $t = 0$  EG: if  $\vec{a} = e^t$ , then  $\vec{v} = e^t + c_1$ . If  $\vec{v}(0) = 0$ , then  $\vec{v}(0) = e^0 + c_1 = 0 \implies c_1 = -1$ .

# Chapter 3

## Partial Derivatives

### 3.1 Functions of Several Variables

We've previously seen calculus defined over single variable vector- and scalar-valued functions, but now we define calculus for scalar functions of multiple variables. We focus on

$$f : \mathbb{R}^2 \rightarrow \mathbb{R}$$

$$f : \mathbb{R}^3 \rightarrow \mathbb{R}$$

$$f : \mathbb{R}^n \rightarrow \mathbb{R}$$

**Definition 3.1.1 (Domain and Range of Multivariate Functions).** If  $f$  is a function of 2 variables, we have

$$f : D \subseteq \mathbb{R} \rightarrow \mathbb{R}$$

where  $D = \text{dom}(f) = \{(x, y) \mid (x, y) \mapsto f(x, y)\}$ .

We also define  $R = \text{rng}(f) = \{f(x, y) \in \mathbb{R} \mid (x, y) \in D\}$ .

We often have  $z = f(x, y)$  to show that  $x, y$  are independent and can vary throughout  $D$ , whereas  $z$  is dependent on  $x, y$ .

Example: The domain of  $f(x, y) = 3 \ln(x + y^2)$  is:

$$D = \{(x, y) \mid x + y^2 > 0\}$$

because the function is defined as  $f : D \rightarrow \mathbb{R}$ , and  $\ln(\cdot) : \mathbb{R}^+ \rightarrow \mathbb{R} \wedge \mathbb{R}^- \rightarrow \mathbb{C}$ , and is undefined for 0. The range will be  $\mathbb{R}$ , obviously.

**Definition 3.1.2 (Level Curve).** For a function  $f(x, y) = z$ , a level curve is the 2d graph formed by taking  $f(x, y = k)$ , where  $k \in \mathbb{R}$  and is not a variable.

We often overlay many level curves onto one 2D graph to form a contour (topological) map. The 4D analogue of a level curve is a level surface formed by taking  $f(x, y, z) = k$  in the same manner.

## 3.2 Limits and Continuity

Let  $f$  be a function defined on an open interval containing  $x$  values arbitrarily close to  $a$ . Recall that

$$\lim_{x \rightarrow a} f(x) = L \iff \forall(\varepsilon > 0)(\exists(\delta > 0)(0 < |x - a| < \delta \implies |f(x) - L| < \varepsilon))$$

We now generalize this to multivariate scalar functions.

**Definition 3.2.1 (Formal Definition of 2D Limit).** Let  $f$  be a function of two variables with domain  $D$  including points arbitrarily close to  $a, b$ .

$$\lim_{(x,y) \rightarrow (a,b)} f(x,y) = L \iff \forall(\varepsilon > 0)(\exists(\delta > 0)(0 < \sqrt{(x-a)^2 + (y-b)^2} < \delta \implies |f(x,y) - L| < \varepsilon))$$

Notice the use of the distance formula, or more generally, the *euclidean norm*. This generalizes to  $n$ -dimensions:

**Definition 3.2.2 (Formal Definition of Limits).** Let  $f$  be a function of  $n$  variables with domain  $D$  including points arbitrarily close to  $\vec{a} \in \mathbb{R}^n$ . For the sake of being concise, let  $\vec{x} = (x_1, \dots, x_n) \in \mathbb{R}^n$ .

$$\lim_{\vec{x} \rightarrow \vec{a}} f(\vec{x}) = L \iff \forall(\varepsilon > 0)(\exists(\delta > 0)(0 < \|\vec{x} - \vec{a}\| < \delta \implies |f(\vec{x}) - L| < \varepsilon))$$

We have the same definition of continuity.

**Definition 3.2.3 (Continuity at a Point).** If  $f(a, b) = \lim_{(x,y) \rightarrow (a,b)} f(x,y)$ , then  $f$  is continuous at  $(a, b)$ .

**Theorem 3.2.1 (Continuous Functions).** All elementary functions (polynomial, logarithmic, radical, rational, exponential, trig, inverse trig) are continuous **on their domains**. Furthermore, compositions of continuous functions are continuous on their domains.

When a function is continuous at a point, we can evaluate its limit at that point by showing that the function is continuous there, and then plugging in values.

**Theorem 3.2.2 (Limit Laws).**

$$\begin{aligned}\lim_{(x,y) \rightarrow (a,b)} x &= a \\ \lim_{(x,y) \rightarrow (a,b)} y &= b \\ \lim_{(x,y) \rightarrow (a,b)} c &= c\end{aligned}$$

**Theorem 3.2.3 (Squeeze Theorem).** If  $f(x) \leq g(x, y) \leq h(x)$  when  $x$  is near  $a$ , and

$$\lim_{(x,y) \rightarrow (a,b)} f(x) = \lim_{(x,y) \rightarrow (a,b)} h(x) = L$$

then

$$\lim_{(x,y) \rightarrow (a,b)} g(x) = L$$

Note the use of single variable functions in the previous theorem.

A good checklist of strategies for finding limits of 2D functions is:

1. Is the function continuous?
2. Can the function be simplified algebraically?
3. Try to show that the limit does *not* exist by showing two different paths  $(x, y) \rightarrow (a, b)$  give different limit values.
4. Try using polar coordinates to find it. Give  $x = r \cos \theta$ ,  $y = r \sin \theta$ , and  $(x, y) \rightarrow (0, 0)$  is equivalent to  $r \rightarrow 0$ . Note this only works for  $(a, b) = (0, 0)$ .
5. Use the squeeze theorem.
6. Use the epsilon-delta definition of limit.

**Example.** Evaluate the limit, or show it does not exist.

$$\lim_{(x,y) \rightarrow (0,0)} \frac{xy^2}{x^2 + y^2}$$

$\frac{0}{0}$  is indeterminate, so  $(0, 0) \notin D$ .

**Path 1:**

$$\begin{aligned} y &= x \text{ and } x \rightarrow 0 \\ \implies \lim_{x \rightarrow 0} \frac{x^3}{x^2 + x^4} &= \lim_{x \rightarrow 0} \frac{x}{1 + x^2} = \frac{0}{1} = 0 \end{aligned}$$

**NOTE:** just because we have shown that one path gives one value does not mean this is the value of the limit. The entire point of this method is to show the same limit has multiple values, thus it can only disprove the existence of a limit, not prove it. Thus, we consider more paths.

**Path 2:**

$$\begin{aligned} x &= y^2 \text{ and } y \rightarrow 0 \\ \lim_{y \rightarrow 0} \frac{y^4}{y^4 + y^4} &= \lim_{y \rightarrow 0} \frac{y^4}{2y^4} = \frac{1}{2} \end{aligned}$$

$0 \neq \frac{1}{2}$ , so the limit does not exist at  $(0, 0)$ .

A note about the previous example: we cannot choose a definition for  $x$  or  $y$  where we do not get to the desired point.

**Example.** Evaluate the limit, or show it does not exist.

$$\lim_{(x,y) \rightarrow (0,0)} \frac{x^2 \sin^2 y}{x^2 + 2y^2}$$

We will use the squeeze theorem.

$$\begin{aligned}
0 &\leq \frac{x^2}{x^2 + 2y^2} \leq 1 \quad \forall x \in \mathbb{R} \\
\implies 0 &\leq \frac{x^2 \sin^2 y}{x^2 + 2y^2} \leq \sin^2 y \\
\lim_{(x,y) \rightarrow (0,0)} 0 &= 0 \\
\lim_{(x,y) \rightarrow (0,0)} \sin^2 y &= 0 \\
\therefore \lim_{(x,y) \rightarrow (0,0)} \frac{x^2 \sin^2 y}{x^2 + 2y^2} &= 0 \text{ by squeeze theorem. } \square
\end{aligned}$$

### 3.3 Partial Derivatives

**Definition 3.3.1 (Partial Derivatives of a Function of 2 Variables).** Let  $f : \mathbb{R}^2 \rightarrow \mathbb{R}$  be the function  $z = f(x, y)$ . Then the partial derivatives are given as

$$\begin{aligned}
f_x(x, y) &= \lim_{h \rightarrow 0} \frac{f(x + h, y) - f(x, y)}{h} \\
f_y(x, y) &= \lim_{h \rightarrow 0} \frac{f(x, y + h) - f(x, y)}{h}
\end{aligned}$$

A way to think about the above definition is by having a function  $g(x) = f(x, b)$ , where  $b$  is constant. In other words, we let  $x$  vary while holding  $y$  constant for some arbitrary  $b \in \mathbb{R}$ . In this case,  $g'(x) = f_x(x, b)$ , or  $g'$  is the partial derivative of  $f$  with respect to  $x$ .

**Notation.** The following notations are equivalent.

$$f_x(x, y) = f_x = \frac{\partial f}{\partial x} = \frac{\partial}{\partial x} f(x, y) = z_x = D_1 f = D_x f$$

Generally we want to avoid big  $D$  notation for partial derivatives. The same statements hold for  $y$ .

**Example.** Let  $f(x, y) = 3x^3y^2 - 4xy$ . Find  $\frac{\partial f}{\partial x}$  and  $\frac{\partial f}{\partial y}$ .

$$\begin{aligned}
\frac{\partial f}{\partial x} &= 9x^2y^2 - 4y \\
\frac{\partial f}{\partial y} &= 6x^3y - 4x
\end{aligned}$$

**Example.** Let  $g(x) = \cos\left(\frac{x^2}{1+y}\right)$ . Find  $g_x$  and  $g_y$ .

$$g_x = -\sin\left(\frac{x^2}{1+y}\right) \left(\frac{2x}{1+y}\right)$$

$$g_y = -\sin\left(\frac{x^2}{1+y}\right) \left(\frac{-x^2}{(1+y)^2}\right)$$

**Remark.** Partial derivatives for functions of  $n$  variables exist in the same way.

**Notation.** Higher order partial derivatives can be given in various ways:

$$(f_x)_y = f_{xy} = \frac{\partial}{\partial y} \left( \frac{\partial f}{\partial x} \right) = \frac{\partial f}{\partial y \partial x} = \frac{\partial}{\partial x} \left( \frac{\partial}{\partial y} (f(x, y)) \right)$$

When in Leibnitz notation, repeated partials are operated from left to right, similarly to function compositions.

**Example.** Let  $f(x, y) = 3x^3y^2 - 4xy$ . From a previous example, we have

$$\frac{\partial f}{\partial x} = 9x^2y^2 - 4y \quad \frac{\partial f}{\partial y} = 6x^3y - 4x$$

Find the second order partials.

$$f_{xx} = 18xy^2 \quad f_{yy} = 6x^3$$

$$f_{xy} = 18x^2y - 4 \quad f_{yx} = 18x^2y - 4$$

**Theorem 3.3.1 (Clairaut's Theorem).** If  $f$  is defined on a disc  $D$  which contains  $(a, b)$ , and if  $f_{xy}$  and  $f_{yx}$  are continuous on  $D$ , then

$$f_{xy}(a, b) = f_{yx}(a, b)$$

**Remark.** The aforementioned disk is assigned the radius of some arbitrary  $\varepsilon$ . From this, a better interpretation of the above theorem can be stated as such:

If  $f$  is continuous on a closed region  $D$ , and its partials are continuous on  $D$ , then the order of differentiation doesn't matter for any points not on the border of  $D$ .

(A disk of radius  $\varepsilon$  is also known as an epsilon ball.)

**Definition 3.3.2 (Laplace's Equation).** The partial differential equation (PDE)  $u_{xx} + u_{yy} = 0$  is known as **Laplace's Equation**. Solutions  $u$  to this equation are known as **harmonic functions**.

**Example.** Does  $u(x, y) = e^x \sin(y)$  satisfy Laplace's Equation?

$$u_{xx} = e^x \sin(y)$$

$$u_{yy} = -e^x \sin(y)$$

$$u_{xx} + u_{yy} = e^x \sin(y) - e^x \sin(y) = 0$$

Therefore this equation is a solution to Laplace's Equation.

**Definition 3.3.3 (Wave Equation).** The **Wave Equation** is a PDE given as  $u_{tt} = a^2 u_{xx}$ .

**Example.** Let  $u(x, t) = \sin(x - 5t)$ . Show  $u$  is a solution to the wave equation.

$$u_t = -5 \cos(x - 5t) \quad u_x = \cos(x - 5t)$$

$$u_{tt} = -25 \sin(x - 5t) \quad u_{xx} = -\sin(x - 5t)$$

$$\implies u_{tt} - 25u_{xx} = -25 \sin(x - 5t) + 25 \sin(x - 5t) = 0$$

$$\implies u_{tt} = 25u_{xx}$$

## 3.4 Tangent Planes and Linear Approximations

**Definition 3.4.1 (Tangent Plane).** If  $z = f(x, y)$  defines a surface with  $P = (x_0, y_0, z_0) \in f$ , then let  $C_1$  and  $C_2$  be the curves  $x = x_0$  and  $y = y_0$ . The tangent plane is the plane containing the lines tangent to  $C_1$  and  $C_2$ .

**Definition 3.4.2 (Linearization).** If  $f(x, y)$  defines a surface with point  $(a, b)$ , then near  $(a, b)$  it follows that

$$f(x, y) \approx L(x, y) = f_x(a, b)(x - a) + f_y(a, b)(y - b)$$

**Definition 3.4.3 (Differentiability of a 2D Function).** Let  $\Delta z = f(a, b) - f(x, y)$ . Let  $\varepsilon_1, \varepsilon_2$  be functions of  $\Delta x$  and  $\Delta y$  such that  $\varepsilon_1, \varepsilon_2$  go to 0 as  $(\Delta x, \Delta y) \rightarrow (0, 0)$ .  $f$  is differentiable at  $(a, b)$  iff

$$\Delta z = f_x(a, b)\Delta x + f_y(a, b)\Delta y + \varepsilon_1\Delta x + \varepsilon_2\Delta y$$

**Definition 3.4.4 (Tangent Plane).** Suppose a function  $z = f(x, y)$  has a point  $(x_0, y_0, z_0) \in f$ . The tangent plane to  $f$  at this point can be written as

$$z - z_0 = f_x(x_0, y_0)(x - x_0) + f_y(x_0, y_0)(y - y_0)$$

**Remark.** The variables in this function are only  $x, y, z$ .



**Corollary 3.4.1.** The normal vector to this plane is

$$\vec{N} = \langle f_x(x_0, y_0), f_y(x_0, y_0), -1 \rangle$$

**Example.** Find the equation of the tangent plane to  $z = y \ln x$  at  $(e, 4, 4)$ .

$$z - 4 = \frac{4}{e}(x - e) + (y - 4)$$

The tangent plane can be used as a **linear approximation** for the function  $z$  near the point  $P$ .

**Definition 3.4.5 (Linearization).** If  $z = f(x, y)$ , then near  $P = (x_0, y_0, z_0) \in f$ ,

$$f(x, y) \approx L(x, y) = f_x(x_0, y_0)(x - x_0) + f_y(x_0, y_0)(y - y_0) + z_0$$

**Theorem 3.4.1.** Let  $f : D \rightarrow \mathbb{R}$  be a function of two variables with  $(a, b) \in D$ . If  $f_x$  and  $f_y$  exist near and at  $(a, b)$ , and are continuous at  $(a, b)$ , then  $f$  is differentiable at  $(a, b)$ .

**Definition 3.4.6 (Differentials).** For a differentiable function  $z = f(x, y)$ , let  $dx$  and  $dy$  be independent variables. Then the **differential** of  $z$  (or **total differential**) is given as

$$dz = f_x(x, y)dx + f_y(x, y)dy$$

## 3.5 The Chain Rule

From Calc 1, we know if  $f$  is a function of a function, that is  $f(g(t)) = f(x)$ , meaning  $x = g(t)$ , then we give

$$f'(x) = f'(g(t))g'(t)$$

Or in Leibnitz' notation,

$$\frac{df}{dt} = \frac{df}{dx} \frac{dx}{dt}$$

This is the chain rule. It's important to realize here that variables can be thought of as nested functions of other variables. This will be how we extend the chain rule to multidimensions.

**Theorem 3.5.1 (Chain Rule, Case 1).** If  $z = f(x, y)$  where  $x = g(t)$  and  $y = h(t)$ , meaning  $z = f(g(t), h(t))$ , then

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

where  $z$  is differentiable.

**Example.** Let  $f(x, y) = x^3y - 4x^2y^2$ , where  $x = \sin(2t)$  and  $y = \cos(3t)$ . Find  $\frac{dz}{dt}$ .

$$\frac{dz}{dt} = (3x^2 - 8xy^2)(2\cos(2t)) + (x^3 - 8x^2y)(-3\sin(3t))$$

**Corollary 3.5.1.** In general, for  $z = f(x_1, \dots, x_n)$  where the partial derivatives are continuous, then

$$\frac{dz}{dt} = \sum_{i=1}^n \frac{\partial z}{\partial x_i} \frac{dx_i}{dt}$$

**Theorem 3.5.2 (Chain Rule, Case 2).** Let  $f(x, y) = z$ ,  $x = g(s, t)$ , and  $y = h(s, t)$ . Then

$$\frac{\partial z}{\partial s} = \frac{\partial z}{\partial x} \frac{\partial x}{\partial s} + \frac{\partial z}{\partial y} \frac{\partial y}{\partial s}$$

Similarly,

$$\frac{\partial z}{\partial t} = \frac{\partial z}{\partial x} \frac{\partial x}{\partial t} + \frac{\partial z}{\partial y} \frac{\partial y}{\partial t}$$

**Theorem 3.5.3 (Chain Rule, General Case).** If  $f$  is a differentiable function in  $n$  variables  $x_1, \dots, x_n$ , and each  $x_i$  is a differentiable function of  $m$  variables  $t_1, \dots, t_m$ , then

$$\begin{aligned} \frac{\partial f}{\partial t_j} &= \frac{\partial f}{\partial x_1} \frac{\partial x_1}{\partial t_j} + \dots + \frac{\partial f}{\partial x_n} \frac{\partial x_n}{\partial t_j} \\ &= \sum_{i=1}^n \frac{\partial f}{\partial x_i} \frac{\partial x_i}{\partial t_j} \end{aligned}$$

We can now do **implicit differentiation** in multiple variables.

**Example.** Let  $w^3 + x^3 + y^3 + z^3 + 6wz = 4y$ . Find  $\frac{\partial z}{\partial x}$ .

$$\frac{\partial}{\partial x} (w^3 + x^3 + y^3 + z^3 + 6wz) = \frac{\partial}{\partial x} (4y)$$

$$3w^2w_x + 3x^2 + 3y^2y_x + 3z^2z_x + 6wz_x + 6zw_x = 4y_x$$

$$3z^2z_x + 6wz_x = 4y_x - 3w^2w_x - 3x^2 - 3y^2y_x - 6zw_x$$

$$z_x(3z^2 + 6w) = 4y_x - 3w^2w_x - 3x^2 - 3y^2y_x - 6zw_x$$

$$z_x = \frac{4y_x - 3w^2w_x - 3x^2 - 3y^2y_x - 6zw_x}{3z^2 + 6w}$$

## 3.6 Directional Derivatives and the Gradient Vector

**Definition 3.6.1 (Gradient Vector).** The gradient vector (field) of a function  $\nabla f$  in  $n$  variables is given as

$$\left\langle \frac{\partial f}{\partial x_1}, \dots, \frac{\partial f}{\partial x_n} \right\rangle$$

**Definition 3.6.2 (Directional Derivative).** The derivative of a function  $f(x, y)$  in the arbitrary direction of some unit vector  $\vec{u} = \langle a, b \rangle$  at a point  $(x_0, y_0)$  is defined as

$$D_{\vec{u}}f = \lim_{h \rightarrow 0} \frac{f(x_0 - ha, y_0 - hb) - f(x_0, y_0)}{h}$$

**Remark.** The above definition can be extended to  $\mathbb{R}^n$ .

**Theorem 3.6.1 (Directional Derivative Thm. 1).** The directional derivative of a function  $f$  in the direction of unit vector  $\vec{u} = \langle a, b \rangle$  is given as

$$D_{\vec{u}}f = f_x(x, y)a + f_y(x, y)b$$

**Remark.** The above definition can be extended to  $\mathbb{R}^n$ .

**Proposition 3.6.1 (Directional Derivative as a Dot Product).** With the previous definitions, we have

$$D_{\vec{u}}f = \nabla f \cdot \vec{u}$$

**Remark.** The gradient vector points in the steepest rate of change and is perpendicular to all level curves (contour map).

**Theorem 3.6.2 (Directional Derivative Thm. 2).** For a differentiable function  $f$  of two variables, the maximum value of  $D_{\vec{u}}f(\vec{x})$  is

$$|\nabla f(\vec{x})|$$

and is in the direction of  $\nabla f$ .

**Remark.** The above definition can be extended to  $\mathbb{R}^n$ .

## 3.7 Maximum and Minimum Values

**Theorem 3.7.1 (First Derivative Test).** If  $f$  is a function of **two variables** that has a local max or min at  $(a, b)$ , and the first order partial derivatives exist at  $(a, b)$ , then

$$\nabla f(a, b) = \vec{0}$$

**Theorem 3.7.2 (Second Derivative Test).** Let  $f$  be a function of **two variables**. Suppose the partial derivatives of  $f$  are continuous on a disk centered at  $(a, b)$ , and  $\nabla f(a, b) = \vec{0}$ . Define a new function  $D$ , the discriminant function:

$$D(x, y) = f_{xx}f_{yy} - (f_{xy})^2$$

- If  $D(a, b) > 0$  and  $f_{xx}$  or  $f_{yy}$  are greater than 0, then  $f(a, b)$  is a local min.
- If  $D(a, b) > 0$  and  $f_{xx}$  or  $f_{yy}$  are less than 0, then  $f(a, b)$  is a local max.
- If  $D(a, b) < 0$ , then  $f(a, b)$  is a saddle point.
- If  $D(a, b) = 0$ , then the test is inconclusive.

**Remark.** A mnemonic for remembering the discriminant function is

$$D = \begin{vmatrix} f_{xx} & f_{xy} \\ f_{yx} & f_{yy} \end{vmatrix}$$

Note that the partials are defined on a disk so 3.3.1 applies, and  $f_{xy}f_{yx} = (f_{xy})^2$ .

**Notation.** This is beyond calc 3.

If  $S$  is a set, then  $\max(S)$  gives the maximum element of the set. Likewise,  $\min(S)$  gives the min.

If a function is defined  $f : X \rightarrow Y$ , then

$$\arg \max_{x \in S} f(x) = \{x \in S \mid \forall s \in X (f(s) \leq f(x))\}$$

$\arg \min$  is defined similarly.

**Theorem 3.7.3 (Extreme Value Theorem for Multivariate Functions).** If  $f$  is defined on a closed, bounded set  $D \subseteq \mathbb{R}^n$ , then  $f$  has an absolute max and min on  $D$ .

To find the extreme values for a function of only **two variables** on a set  $D \subseteq \mathbb{R}^2$ , set  $\nabla f = \vec{0}$ , solve for points, evaluate  $f$  at the points found, and evaluate  $f$  on the boundary of  $D$  (notated  $\partial D$ ).

## 3.8 Lagrange Multipliers

The method of Lagrange multipliers can optimize a function given some constraint. Suppose we have some function  $f$  and a constraint function  $g$ , where  $g$  has some set value. Find values where  $\nabla f \parallel \lambda \nabla g$ . Set up and solve a system of equations. Plug in values to the original function.

**Example.** Let  $f(x, y) := (x - 2)^2 + (y - 2)^2$  and  $g(x, y) := x^2 + y^2 = 9$ .

$$\nabla f = \langle 2x - 4, 2y - 4 \rangle$$

$$\nabla g = \langle 2x, 2y \rangle$$

$$\nabla f = \lambda \nabla g$$

$$\implies \begin{cases} 2x - 4 = \lambda 2x \\ 2y - 4 = \lambda 2y \end{cases}$$

Solve system of equations

$$x, y \in \left\{ 1 + \frac{2\sqrt{2}}{3}, 1 - \frac{2\sqrt{2}}{3} \right\}$$

Plug in values to  $f$  to find the maxes and mins.

# Chapter 4

## Multiple Integrals

### 4.1 Double Integrals over Rectangles

Let a function  $f : \mathbb{R}^2 \rightarrow \mathbb{R}$  be continuous on  $D \subset \mathbb{R}^2$ , where

$$D = [a, b] \times [c, d]$$

Using Riemann sums, we can extend the definition of an integral over rectangular regions such as this. Let  $[a, b]$  be partitioned into  $n$  disjoint subsets of an equal length  $\Delta x$ , and let  $[c, d]$  be similarly partitioned into  $m$  disjoint subsets of an equal length  $\Delta y$ . Let  $(x_i^*, y_j^*)$  be the midpoint of the  $ij$ th subset of  $[a, b] \times [c, d]$ . We say that the volume under some function  $f(x, y)$  over the region  $D$  is approximately

$$V \approx \sum_{i=1}^n \sum_{j=1}^m f(x_i^*, y_j^*) \Delta y \Delta x$$

We say that  $\Delta y \Delta x = \Delta A$  for convenience. This leads to the following definition.

**Definition 4.1.1 (Double Integral over a Rectangular Region).** Let  $f$  be defined over a rectangular region  $R \subset \mathbb{R}^2$ . We say

$$\iint_R f(x, y) dA = \lim_{(n,m) \rightarrow \infty} \sum_{i=1}^n \sum_{j=1}^m f(x_i^*, y_j^*) \Delta A$$

provided the limit exists.

**Remark.** When we integrate a function of multiple variables with respect to a single variable, we hold other variables fixed with respect to the variable being integrated with respect to. This is known as **partial integration**.

**Theorem 4.1.1 (Fubini's Theorem).** Let  $f$  be continuous on  $D = \{(x, y) \mid x \in [a, b] \wedge y \in [c, d]\}$ .

$[c, d]\} \subset \mathbb{R}^2$ . Then

$$\iint_D f(x, y) dA = \int_a^b \int_c^d f(x, y) dy dx = \int_c^d \int_a^b f(x, y) dx dy$$

**Remark.** The restriction of continuity can be relaxed in some cases, and it can be required only that  $f$  be bounded on  $D$  and be discontinuous on a finite number of smooth curves. Proof of this fact most likely follows from utilization of lebesgue integration, but either way it'll likely be shown in an analysis course.

## 4.2 Double Integrals over General Regions

**Definition 4.2.1 (Type I Region).** A type I region is a region bounded by two continuous functions of  $x$ .

**Definition 4.2.2 (Type II Region).** A type II region is a region bounded by two continuous functions of  $y$ .

**Definition 4.2.3 (Integrals over Type I or II Regions).** Let  $D_1$  be a type I region with

$$D_1 = \{(x, y) \in \mathbb{R}^2 \mid x \in [a, b] \wedge y \in [g_1(x), g_2(x)]\}$$

and let  $f$  be defined and integrable over  $D_1$ . Then

$$\iint_{D_1} f(x, y) dA = \int_a^b \int_{g_1(x)}^{g_2(x)} f(x, y) dy dx$$

Furthermore, let  $D_2$  be a type II region with

$$D_2 = \{(x, y) \in \mathbb{R}^2 \mid x \in [h_1(y), h_2(y)] \wedge y \in [c, d]\}$$

and let  $f$  be defined and integrable over  $D_2$ . Then

$$\iint_{D_2} f(x, y) dA = \int_c^d \int_{h_1(y)}^{h_2(y)} f(x, y) dx dy$$

**Remark.** Notice that when integrating over a type I or type II region, the integral bounded by constants is the last integral evaluated.

The following are properties of double integrals.

1. If  $f(x, y) \geq 0 \forall (x, y) \in D$ , then the volume  $V$  of the solid region above  $D$  and below  $z = f(x, y)$

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is

$$V = \iint_D f(x, y) dA$$

2. The average value of a function  $f : \mathbb{R}^2 \rightarrow \mathbb{R}$  on  $D \subset \mathbb{R}^2$  is

$$f_{avg} = \frac{1}{A(D)} \iint_D f(x, y) dA$$

where  $A(D)$  is the area of the region  $D$ .

3. Integral of a sum:

$$\iint_D f(x, y) + g(x, y) dA = \iint_D f(x, y) dA + \iint_D g(x, y) dA$$

4. Integral of a constant multiple:

$$\iint_D c f(x, y) dA = c \iint_D f(x, y) dA$$

5. If  $f(x, y) \geq g(x, y) \forall (x, y) \in D$ , then

$$\iint_D f(x, y) dA \geq \iint_D g(x, y) dA$$

6. If  $D = \bigcup_{i=1}^n D_i$  where  $\bigcap_{i=1}^n D_i = \emptyset$ , then

$$\iint_D f(x, y) dA = \sum_{i=1}^n \iint_{D_i} f(x, y) dA$$

7. Integral of the constant function:

$$\iint_D 1 dA = A(D)$$

8. If  $m \leq f(x, y) \leq M \forall (x, y) \in D$ , then

$$mA(D) \leq \iint_D f(x, y) dA \leq MA(D)$$

## 4.3 Double Integrals in Polar Coordinates

Integrals over polar coordinates are defined through a Riemann sum of polar rectangles. It's very convoluted so I will leave the proof as an exercise to myself when I feel like it in the future.



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**Definition 4.3.1 (Polar Rectangle).** A set of the form  $\{(r \cos \theta, r \sin \theta) \mid r \in [a, b] \wedge \theta \in [\alpha, \beta]\}$ . This can be seen as a subset of a circle in both radius and angle.

**Theorem 4.3.1 (Polar Integrals).** If  $f$  is continuous on a polar rectangle  $R$  with  $0 \leq a \leq r \leq b$  and  $\alpha \leq \theta \leq \beta$ , where  $\beta - \alpha \leq 2\pi$ , then

$$\iint_R f(x, y) dA = \int_{\alpha}^{\beta} \int_a^b f(r \cos \theta, r \sin \theta) r dr d\theta$$

Notice the extra  $r$  in the integrand in the above theorem.