

# Multivariable Calculus

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**Definition 0.0.1: Common Notation**

The notes will be using common mathematical notations.

$\mathbb{Z}$	The set of all integers
$\mathbb{Q}$	The set of all rational numbers
$\mathbb{R}$	The set of all real numbers
$\mathbb{C}$	The set of all complex numbers
$\{x_1, \dots, x_n\}$	Finite set with elements $x_1, x_2, x_3, \dots, x_n$
$x \in E$	$x$ is an element of set $E$
$x \notin E$	$x$ is not an element of set $E$
$A \subset B$	Set $A$ is a subset set $E$
$A \not\subset B$	Set $A$ is not a subset set $E$
$f: A \rightarrow B$	Function $f$ maps set $A$ into set $B$

To note, most theorems will include proofs, but some are beyond the scope of this course and thus, not included. However, none of the proofs are required in order for these theorems to be applied.

# 1 Vectors

## 1.1 Vectors

**Definition 1.1.1: Vectors**

A **scalar**  $c$  is a number in  $\mathbb{R}$ .

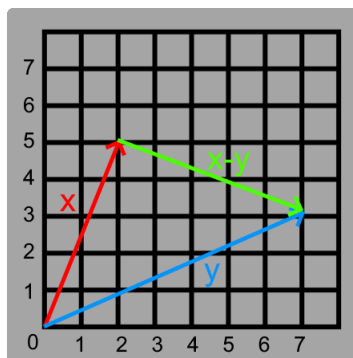
A **vector**  $x \in \mathbb{R}^n$  is an ordered  $n$ -tuple of real numbers.

$$x = (x_1, \dots, x_n) = \langle x_1, \dots, x_n \rangle \quad \text{where each } x_i \in \mathbb{R}$$

Let the zero vector  $0 = (0, \dots, 0)$ .

If  $x, y \in \mathbb{R}^n$  and  $c$  is a scalar:

Comparison:	$x = y$ if $x_i = y_i$ for $i = \{1, \dots, n\}$
Vector Addition:	$x + y = (x_1 + y_1, \dots, x_n + y_n)$
Scalar Multiplication:	$cx = (cx_1, \dots, cx_n)$

**Theorem 1.1.2: Vector Operations**

(a)  $x + y = y + x$

**Proof**

$$x + y = (x_1 + y_1, \dots, x_n + y_n) = (y_1 + x_1, \dots, y_n + x_n) = y + x$$

(b)  $x + (y + z) = (x + y) + z$

**Proof**

$$\begin{aligned} x + (y + z) &= (x_1, \dots, x_n) + (y_1 + z_1, \dots, y_n + z_n) = (x_1 + y_1 + z_1, \dots, x_n + y_n + z_n) \\ &= (x_1 + y_1, \dots, x_n + y_n) + (z_1, \dots, z_n) = (x + y) + z \end{aligned}$$

(c)  $x+0 = x$

**Proof**

$$x+0 = (x_1 + 0, \dots, x_n + 0) = (x_1, \dots, x_n) = x$$

(d)  $c(x+y) = cx + cy$

**Proof**

$$c(x+y) = (c(x_1 + y_1), \dots, c(x_n + y_n)) = (cx_1 + cy_1, \dots, cx_n + cy_n) = cx + cy$$

(e)  $(c+k)v = cv + kv$

**Proof**

$$(c+k)v = ((c+k)v_1, \dots, (c+k)v_n) = (cv_1 + kv_1, \dots, cv_n + kv_n) = cv + kv$$

(f)  $c(kx) = (ck)x = k(cx)$

**Proof**

$$\begin{aligned} c(kv) &= (c(kx_1), \dots, c(kx_n)) = (ckx_1, \dots, ckx_n) = (ck)x = (kcx_1, \dots, kcx_n) \\ &= (k(cx_1), \dots, k(cx_n)) = k(cx) \end{aligned}$$

**Definition 1.1.3: Standard Basis Vectors**

The **standard basis vectors** for  $\mathbb{R}^n$  are  $e_1, \dots, e_n$  where each  $i = \{1, \dots, n\}$ :

$$e_i = (0, 0, \dots, 0, 1, 0, \dots, 0, 0, \dots)$$

1    2            i-1    i    i+1            n-1    n

Thus, for any  $x \in \mathbb{R}^n$ :

$$x = (x_1, \dots, x_n) = x_1 e_1 + \dots + x_n e_n$$

**Example**

Let  $x = (4, -2, 5)$  and  $y = (-7, 1, 2)$ . Find  $3x-2y$  in standard basis vectors.

$$3x+y = 3 \begin{bmatrix} 4 \\ -2 \\ 5 \end{bmatrix} - 2 \begin{bmatrix} -7 \\ 1 \\ 2 \end{bmatrix} = \begin{bmatrix} 12 \\ -6 \\ 15 \end{bmatrix} - \begin{bmatrix} -14 \\ 2 \\ 4 \end{bmatrix} = \begin{bmatrix} 26 \\ -8 \\ 11 \end{bmatrix} = 26 \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} - 8 \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} + 11 \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$$

**1.2 Dot Product****Definition 1.2.1: Dot Product, Norm, and Orthogonality**

The **dot product** of  $x, y \in \mathbb{R}^n$  is the sum of the products of their components:

$$x \cdot y = x_1 y_1 + \dots + x_n y_n$$

The length of  $x \in \mathbb{R}^n$  is the **norm**:

$$\|x\| = \sqrt{x_1^2 + \dots + x_n^2} = \sqrt{x \cdot x} \quad \Rightarrow \quad x \cdot x = \|x\|^2$$

Thus,  $\|cx\| = \sqrt{(cx_1)^2 + \dots + (cx_n)^2} = |c| \sqrt{x_1^2 + \dots + x_n^2} = |c| \|x\|$ .

Then, a **unit vector** (i.e. vector of length 1) in the direction of  $x$  is  $\frac{x}{\|x\|}$ .

$x, y \in \mathbb{R}^n$  are **orthogonal** (i.e. perpendicular) if:

$$x \cdot y = 0$$

**Theorem 1.2.2: Properties of the Dot Product**

(a)  $x \cdot x \geq 0$

Proof

$$x \cdot x = x_1x_1 + \dots + x_nx_n = x_1^2 + \dots + x_n^2 \geq 0 + \dots + 0 = 0$$

(b)  $x \cdot x = 0$  if and only if  $x = 0$

Proof

$$x \cdot x = x_1x_1 + \dots + x_nx_n = x_1^2 + \dots + x_n^2$$

Thus,  $x \cdot x = 0$  if and only if each  $x_i^2 = 0$  so each  $x_i = 0$ . Thus,  $x = 0$ .

(c)  $x \cdot y = y \cdot x$

Proof

$$x \cdot y = x_1y_1 + \dots + x_ny_n = y_1x_1 + \dots + y_nx_n = y \cdot x$$

(d)  $x \cdot (y + z) = x \cdot y + x \cdot z$

Proof

$$\begin{aligned} x \cdot (y + z) &= x_1(y_1 + z_1) + \dots + x_n(y_n + z_n) = (x_1y_1 + x_1z_1) + \dots + (x_ny_n + x_nz_n) \\ &= (x_1y_1 + \dots + x_ny_n) + (x_1z_1 + \dots + x_nz_n) = x \cdot y + x \cdot z \end{aligned}$$

(e)  $(x + y) \cdot z = x \cdot z + y \cdot z$

Proof

$$\begin{aligned} (x + y) \cdot z &= (x_1 + y_1)z_1 + \dots + (x_n + y_n)z_n = (x_1z_1 + y_1z_1) + \dots + (x_nz_n + y_nz_n) \\ &= (x_1z_1 + \dots + x_nz_n) + (y_1z_1 + \dots + y_nz_n) = x \cdot z + y \cdot z \end{aligned}$$

(f)  $cx \cdot y = c(x \cdot y) = x \cdot cy$

Proof

$$\begin{aligned} cx \cdot y &= (cx_1)y_1 + \dots + (cx_n)y_n = c(x_1y_1) + \dots + c(x_ny_n) = c(x \cdot y) \\ &= x_1(cy_1) + \dots + x_n(cy_n) = x \cdot cy \end{aligned}$$

**Theorem 1.2.3:  $x \cdot y = \|x\| \|y\| \cos(\theta)$** For  $x, y \in \mathbb{R}^n$ :

$$x \cdot y = \|x\| \|y\| \cos(\theta)$$

where  $\theta \in [0, \pi]$  is the angle between  $x$  and  $y$ ProofSince  $x$ ,  $y$ , and  $x - y$  form a triangle, by the Law of Cosine:

$$\|x - y\|^2 = \|x\|^2 + \|y\|^2 - 2\|x\| \|y\| \cos(\theta)$$

where  $\theta \in [0, \pi]$  is the angle between  $x$  and  $y$ . Since:

$$\|x - y\|^2 = (x - y) \cdot (x - y) = x \cdot x + y \cdot y - 2(x \cdot y) = \|x\|^2 + \|y\|^2 - 2(x \cdot y)$$

then  $x \cdot y = \|x\| \|y\| \cos(\theta)$ .**Example**Let  $x = (1, 2, 2)$  and  $y = (-3, 4, 12)$ . Find the angle between  $x$  and  $y$ .

$$\begin{aligned} 1^2 + 2^2 + 2^2 - 2\sqrt{1^2 + 2^2 + 2^2}\sqrt{(-3)^2 + 4^2 + 12^2} \cos(\theta) \\ 29 = 3 \cdot 13 \cos(\theta) \quad \Rightarrow \quad \theta \approx 0.73 \text{ radians} \end{aligned}$$

**Theorem 1.2.4: Vector Projection**

The projection of  $x \in \mathbb{R}^n$  onto  $y \in \mathbb{R}^n$  is the component of  $x$  parallel to  $y$ :

$$\text{proj}_y x = \frac{x \cdot y}{\|y\|^2} y$$

**Proof**

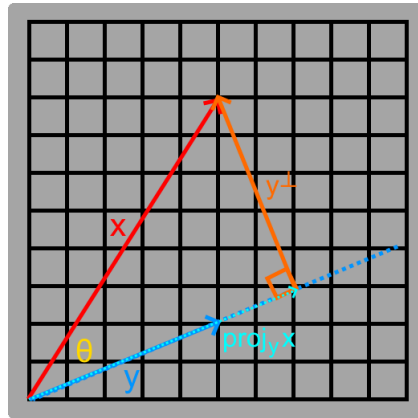
Since  $\text{proj}_y x$  is parallel to  $y$ , let  $\text{proj}_y x = cy$  for some constant  $c \in \mathbb{R}$ .

Let  $y^\perp$  be the orthogonal component of  $x$  to  $y$ . Thus,  $x = \text{proj}_y x + y^\perp = cy + y^\perp$ .

Since  $y^\perp$  is orthogonal to  $y$ , then:

$$x \cdot y = (cy + y^\perp) \cdot y = cy \cdot y + y^\perp \cdot y = cy \cdot y = c\|y\|^2$$

Thus,  $c = \frac{x \cdot y}{\|y\|^2}$  so  $\text{proj}_y x = cy = \frac{x \cdot y}{\|y\|^2} y$ .

**Example**

Let  $x = (-4, 5, 7)$  and  $y = (2, -4, 1)$ . Find the vector decomposition of  $x$  onto  $y$ .

Parallel component:  $\text{proj}_y x = \frac{(-4 \cdot 2 + 5 \cdot (-4) + 7 \cdot 1)}{2^2 + (-4)^2 + 1^2} (2, -4, 1) = \frac{-21}{21} (2, -4, 1) = (-2, 4, -1)$

Orthogonal component:  $x - \text{proj}_y x = (-4, 5, 7) - (-2, 4, -1) = (-2, 1, 8)$

**Theorem 1.2.5: Cauchy-Schwarz Inequality**

For  $x, y \in \mathbb{R}^n$ ,  $|x \cdot y| \leq \|x\| \|y\|$

**Proof**

Let  $y = \text{proj}_x y + x^\perp = cy + x^\perp$  where  $x^\perp$  is the orthogonal component of  $y$  to  $x$  and  $\text{proj}_x y = cy$  is the parallel component to  $x$  for some  $c \in \mathbb{R}$ .

$$x \cdot y = x \cdot (cy + x^\perp) = c(x \cdot x) + x \cdot x^\perp = c\|x\|^2 + 0 = c\|x\|^2$$

Thus,  $c = \frac{x \cdot y}{\|x\|^2}$ . Then:

$$\begin{aligned} \|y\|^2 &= \|cy + x^\perp\|^2 = (cy + x^\perp) \cdot (cy + x^\perp) = cy \cdot cy + x^\perp \cdot x^\perp + 2(cy \cdot x^\perp) \\ &= c^2\|x\|^2 + \|x^\perp\|^2 = \left(\frac{x \cdot y}{\|x\|^2}\right)^2 \|x\|^2 + \|x^\perp\|^2 \end{aligned}$$

$$\|x\|^2 \|y\|^2 = \|x\|^2 \left( \left(\frac{x \cdot y}{\|x\|^2}\right)^2 \|x\|^2 + \|x^\perp\|^2 \right) = (x \cdot y)^2 + \|x\|^2 \|x^\perp\|^2$$

Since  $\|x\|^2 \|x^\perp\|^2 \geq 0$ , then  $(x \cdot y)^2 \leq \|x\|^2 \|y\|^2$  so  $|x \cdot y| \leq \|x\| \|y\|$ .

**Theorem 1.2.6: Triangle Inequality**

For  $x, y \in \mathbb{R}^n$ ,  $\|x + y\| \leq \|x\| + \|y\|$

**Proof**

$$\begin{aligned} \|x + y\|^2 &= (x + y) \cdot (x + y) = x \cdot x + y \cdot y + 2(x \cdot y) = \|x\|^2 + \|y\|^2 + 2(x \cdot y) \\ &\leq \|x\|^2 + \|y\|^2 + 2\|x\| \|y\| \leq \|x\|^2 + \|y\|^2 + 2\|x\| \|y\| = (\|x\| + \|y\|)^2 \end{aligned}$$

## 1.3 Cross Product

### Definition 1.3.1: Cross Product

The **cross product** of  $x, y \in \mathbb{R}^3$  is the determinant of the standard basis,  $x, y$ :

$$x \times y = \det \begin{pmatrix} e_1 & e_2 & e_3 \\ x_1 & x_2 & x_3 \\ y_1 & y_2 & y_3 \end{pmatrix} = \begin{vmatrix} x_2 & x_3 \\ y_2 & y_3 \end{vmatrix} e_1 - \begin{vmatrix} x_1 & x_3 \\ y_1 & y_3 \end{vmatrix} e_2 + \begin{vmatrix} x_1 & x_2 \\ y_1 & y_2 \end{vmatrix} e_3$$

### Theorem 1.3.2: Properties of the Cross Product

(a)  $x \times y = -(y \times x)$

**Proof**

$$\begin{aligned} x \times y &= \begin{vmatrix} x_2 & x_3 \\ y_2 & y_3 \end{vmatrix} e_1 - \begin{vmatrix} x_1 & x_3 \\ y_1 & y_3 \end{vmatrix} e_2 + \begin{vmatrix} x_1 & x_2 \\ y_1 & y_2 \end{vmatrix} e_3 \\ &= - \begin{vmatrix} y_2 & y_3 \\ x_2 & x_3 \end{vmatrix} e_1 + \begin{vmatrix} y_1 & y_3 \\ x_1 & x_3 \end{vmatrix} e_2 - \begin{vmatrix} y_1 & y_2 \\ x_1 & x_2 \end{vmatrix} e_3 = -(y \times x) \end{aligned}$$

(b)  $x \times (y + z) = x \times y + x \times z$

**Proof**

$$\begin{aligned} x \times (y + z) &= \begin{vmatrix} x_2 & x_3 \\ y_2 + z_2 & y_3 + z_3 \end{vmatrix} e_1 - \begin{vmatrix} x_1 & x_3 \\ y_1 + z_1 & y_3 + z_3 \end{vmatrix} e_2 + \begin{vmatrix} x_1 & x_2 \\ y_1 + z_1 & y_2 + z_2 \end{vmatrix} e_3 \\ &= \left( \begin{vmatrix} x_2 & x_3 \\ y_2 & y_3 \end{vmatrix} + \begin{vmatrix} x_2 & x_3 \\ z_2 & z_3 \end{vmatrix} \right) e_1 - \left( \begin{vmatrix} x_1 & x_3 \\ y_1 & y_3 \end{vmatrix} + \begin{vmatrix} x_1 & x_3 \\ z_1 & z_3 \end{vmatrix} \right) e_2 + \left( \begin{vmatrix} x_1 & x_2 \\ y_1 & y_2 \end{vmatrix} + \begin{vmatrix} x_1 & x_2 \\ z_1 & z_2 \end{vmatrix} \right) e_3 \\ &= \begin{vmatrix} x_2 & x_3 \\ y_2 & y_3 \end{vmatrix} e_1 - \begin{vmatrix} x_1 & x_3 \\ y_1 & y_3 \end{vmatrix} e_2 + \begin{vmatrix} x_1 & x_2 \\ y_1 & y_2 \end{vmatrix} e_3 + \begin{vmatrix} x_2 & x_3 \\ z_2 & z_3 \end{vmatrix} e_1 - \begin{vmatrix} x_1 & x_3 \\ z_1 & z_3 \end{vmatrix} e_2 + \begin{vmatrix} x_1 & x_2 \\ z_1 & z_2 \end{vmatrix} e_3 \\ &= x \times y + x \times z \end{aligned}$$

(c)  $(x + y) \times z = x \times z + y \times z$

**Proof**

$$\begin{aligned} (x + y) \times z &= -[z \times (x + y)] = -[z \times x + z \times y] \\ &= -[-(x \times z) + -(y \times z)] = x \times z + y \times z \end{aligned}$$

(d)  $c(x \times y) = cx \times y = x \times cy$

**Proof**

$$\begin{aligned} c(x \times y) &= c \begin{vmatrix} x_2 & x_3 \\ y_2 & y_3 \end{vmatrix} e_1 - c \begin{vmatrix} x_1 & x_3 \\ y_1 & y_3 \end{vmatrix} e_2 + c \begin{vmatrix} x_1 & x_2 \\ y_1 & y_2 \end{vmatrix} e_3 \\ &= \begin{vmatrix} cx_2 & cx_3 \\ y_2 & y_3 \end{vmatrix} e_1 - \begin{vmatrix} cx_1 & cx_3 \\ y_1 & y_3 \end{vmatrix} e_2 + \begin{vmatrix} cx_1 & cx_2 \\ y_1 & y_2 \end{vmatrix} e_3 = cx \times y \\ &= \begin{vmatrix} x_2 & x_3 \\ cy_2 & cy_3 \end{vmatrix} e_1 - \begin{vmatrix} x_1 & x_3 \\ cy_1 & cy_3 \end{vmatrix} e_2 + \begin{vmatrix} x_1 & x_2 \\ cy_1 & cy_2 \end{vmatrix} e_3 = x \times cy \end{aligned}$$

(e)  $x \times x = 0$

**Proof**

$$\text{By part (a), } x \times x = -(x \times x) \quad \Rightarrow \quad 2(x \times x) = 0 \quad \Rightarrow \quad x \times x = 0$$

**Theorem 1.3.3: Orthogonality of  $x \times y$**  $x \times y$  is orthogonal to  $x$  and  $y$ **Proof**

$$\begin{aligned}
x \times y \cdot x &= \left( \begin{vmatrix} x_2 & x_3 \\ y_2 & y_3 \end{vmatrix}, -\begin{vmatrix} x_1 & x_3 \\ y_1 & y_3 \end{vmatrix}, \begin{vmatrix} x_1 & x_2 \\ y_1 & y_2 \end{vmatrix} \right) \cdot (x_1, x_2, x_3) \\
&= (x_2y_3 - x_3y_2)x_1 - (x_1y_3 - x_3y_1)x_2 + (x_1y_2 - x_2y_1)x_3 \\
&= x_1x_2y_3 - x_1x_3y_2 - x_1x_2y_3 + x_2x_3y_1 + x_1x_3y_2 - x_2x_3y_1 = 0 \\
x \times y \cdot y &= \left( \begin{vmatrix} x_2 & x_3 \\ y_2 & y_3 \end{vmatrix}, -\begin{vmatrix} x_1 & x_3 \\ y_1 & y_3 \end{vmatrix}, \begin{vmatrix} x_1 & x_2 \\ y_1 & y_2 \end{vmatrix} \right) \cdot (y_1, y_2, y_3) \\
&= (x_2y_3 - x_3y_2)y_1 - (x_1y_3 - x_3y_1)y_2 + (x_1y_2 - x_2y_1)y_3 \\
&= x_2y_1y_3 - x_3y_1y_2 - x_1y_2y_3 + x_3y_1y_2 + x_1y_2y_3 - x_2y_1y_3 = 0
\end{aligned}$$

**Example**Let  $x = (1, 0, -3)$  and  $y = (6, 8, 2)$ . Find an orthogonal vector to  $x$  and  $y$ .

$$x \times y = \begin{vmatrix} e_1 & e_2 & e_3 \\ 1 & 0 & -3 \\ 6 & 8 & 2 \end{vmatrix} = ((0 \cdot 2 - 8 \cdot -3), (1 \cdot 2 - 6 \cdot -3), (1 \cdot 8 - 6 \cdot -0)) = (24, -20, 8)$$

**Theorem 1.3.4:  $\|x \times y\| = \|x\| \|y\| \sin(\theta)$** For  $x, y \in \mathbb{R}^3$ :

$$\|x \times y\| = \|x\| \|y\| \sin(\theta)$$

where  $\theta \in [0, \pi]$  is the angle between  $x$  and  $y$ **Proof**By **theorem 1.2.3**,  $x \cdot y = \|x\| \|y\| \cos(\theta)$  where  $\theta \in [0, \pi]$  is the angle between  $x, y$ .

$$\|x\|^2 \|y\|^2 - (x \cdot y)^2 = \|x\|^2 \|y\|^2 (1 - \cos^2(\theta)) = \|x\|^2 \|y\|^2 \sin^2(\theta)$$

Also:

$$\begin{aligned}
\|x\|^2 \|y\|^2 - (x \cdot y)^2 &= (x_1^2 + x_2^2 + x_3^2)(y_1^2 + y_2^2 + y_3^2) - (x_1y_1 + x_2y_2 + x_3y_3)^2 \\
&= (x_1^2y_1^2 + x_2^2y_2^2 + x_3^2y_3^2 + \textcolor{teal}{x_1^2y_2^2} + \textcolor{teal}{x_1^2y_3^2} + \textcolor{teal}{x_2^2y_1^2} + \textcolor{teal}{x_2^2y_3^2} + \textcolor{teal}{x_3^2y_1^2} + \textcolor{teal}{x_3^2y_2^2} \\
&\quad - x_1^2y_1^2 - x_2^2y_2^2 - x_3^2y_3^2 - \textcolor{teal}{2x_1x_2y_1y_2} - \textcolor{teal}{2x_1x_3y_1y_3} - \textcolor{teal}{2x_2x_3y_2y_3}) \\
&= (\textcolor{teal}{x_2y_3} - \textcolor{teal}{x_3y_2})^2 + (\textcolor{teal}{x_3y_1} - \textcolor{teal}{x_1y_3})^2 + (\textcolor{teal}{x_1y_2} - \textcolor{teal}{x_2y_1})^2 = \|x \times y\|^2
\end{aligned}$$

Thus,  $\|x \times y\| = \|x\| \|y\| \sin(\theta)$ .



**Theorem 1.3.5: Area of Parallelogram**

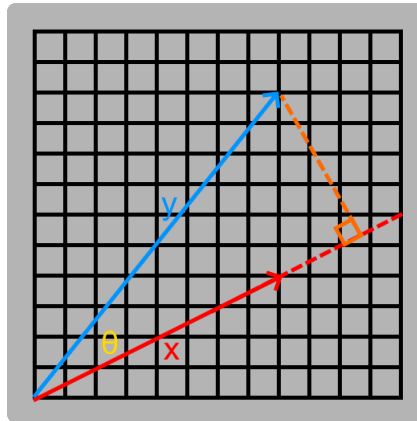
The area of a parallelogram  $P$  with sides  $x, y \in \mathbb{R}^3$ :

$$\text{Vol}_2(P(x, y)) = \|x \times y\|$$

**Proof**

Since parallelogram  $P$  with sides  $x$  and  $y$  is two triangles with sides  $x$  and  $y$ , then:

$$\begin{aligned} \text{Vol}_2(P(x, y)) &= 2 * \text{Vol}_2(\text{Triangle}(x, y)) \\ &= 2 * \frac{1}{2} (\text{base of triangle}) * (\text{height of triangle}) \\ &= \|x\| * (\|y\| \sin(\theta)) = \|x \times y\| \end{aligned}$$

**Theorem 1.3.6: Volume of Parallelepiped**

The volume of a parallelepiped  $P$  with sides  $x, y, z \in \mathbb{R}^n$ :

$$\text{Vol}_3(P(x, y, z)) = |(x \times y) \cdot z|$$

**Proof**

Let sides  $x$  and  $y$  form a base for  $P$ .

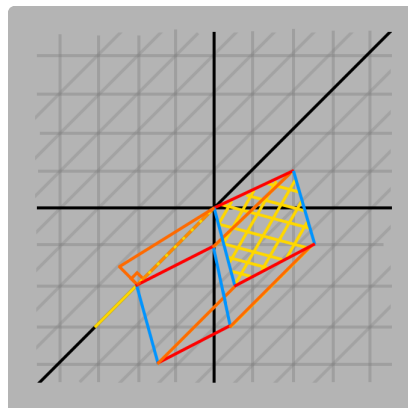
$$\text{Vol}_3(P(x, y, z)) = (\text{Area of base}) * (\text{height}) = \|x \times y\| * (\|z\| \cos(\theta))$$

where  $\theta \in [0, \pi]$  is the angle between  $x \times y$  and  $z$ . By **theorem 1.2.3**:

$$\text{Vol}_3(P(x, y, z)) = (x \times y) \cdot z$$

Since  $-1 \leq \cos(\theta) \leq 1$  for  $\theta \in [0, 2\pi]$ , then  $(x \times y) \cdot z$  can be negative. Thus:

$$\text{Vol}_3(P(x, y, z)) = |(x \times y) \cdot z|$$



## 1.4 Distances and Planes

### Theorem 1.4.1: Equation of a Plane: Method #1: Point and Normal Vector

A plane in  $\mathbb{R}^3$  through a point  $p = (p_x, p_y, p_z)$  and orthogonal to a vector called a normal vector  $n = (a, b, c)$  has an equation of the form:

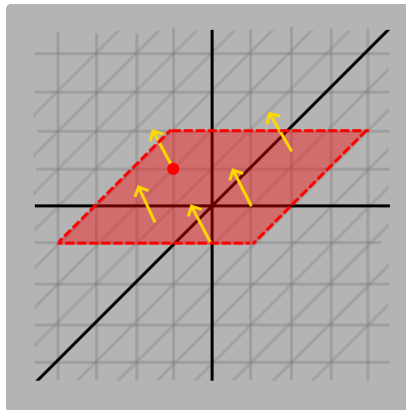
$$n \cdot [(x, y, z) - p] = a(x - p_x) + b(y - p_y) + c(z - p_z) = 0$$

#### Proof

Let  $(x, y, z)$  be any point in the plane. Then  $(x, y, z) - p = (x - p_x, y - p_y, z - p_z)$  is a vector parallel to the plane. Since the plane is orthogonal to vector  $n$ , then any vector parallel to the plane is orthogonal to  $n$ . Thus:

$$n \cdot (x - p_x, y - p_y, z - p_z) = 0$$

$$a(x - p_x) + b(y - p_y) + c(z - p_z) = 0$$



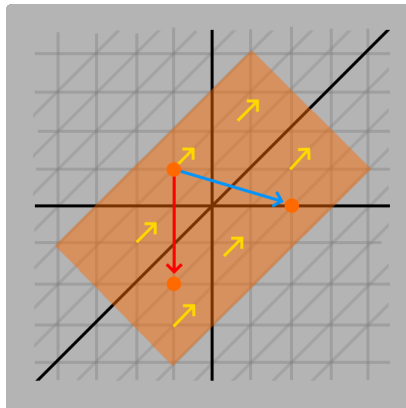
### Theorem 1.4.2: Equation of a Plane: Method #2: 3 Points

A plane in  $\mathbb{R}^3$  through points  $p_1 = (x_1, y_1, z_1)$ ,  $p_2 = (x_2, y_2, z_2)$ , and  $p_3 = (x_3, y_3, z_3)$  has an equation of the form:

$$[(p_2 - p_1) \times (p_3 - p_1)] \cdot [(x, y, z) - p_1] = 0$$

#### Proof

Since  $p_1$ ,  $p_2$ , and  $p_3$  are on the plane, then  $p_2 - p_1$  and  $p_3 - p_1$  are vectors on the plane and thus, parallel to the plane. Since  $(p_2 - p_1) \times (p_3 - p_1)$  is orthogonal to  $(p_2 - p_1)$  and  $(p_3 - p_1)$ , then  $(p_2 - p_1) \times (p_3 - p_1)$  is orthogonal to the plane and thus, a normal vector. By [theorem 1.4.1](#), then  $[(p_2 - p_1) \times (p_3 - p_1)] \cdot [(x, y, z) - p_1] = 0$ .



**Theorem 1.4.3: Distance: Point + Line or 2 Parallel Lines**

The distance from line  $L(t) = tv + x_0$  to point  $p \in \mathbb{R}^3$  where  $t \in \mathbb{R}$ ,  $v, x_0 \in \mathbb{R}^3$ :

$$\frac{\|v \times (p - x_0)\|}{\|v\|}$$

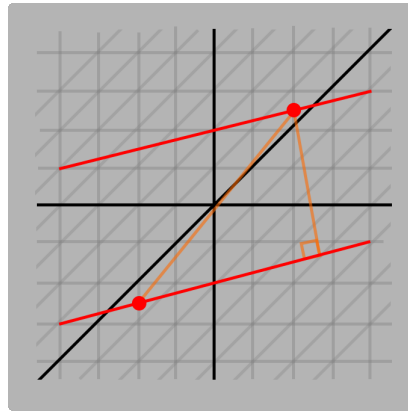
If line  $L_2(t)$  is parallel to  $L(t)$ , choose a point on  $L_2(t)$  and apply formula above to get the distance between two parallel lines.

**Proof**

Since  $x_0$  is a point on  $L(t)$ , then  $p - x_0$  is a vector from line  $L(t)$  to  $p$ .

Let  $\theta$  be the angle between  $p - x_0$  and  $L(t)$ . Thus:

$$\sin(\theta) = \frac{d}{\|p - x_0\|} \Rightarrow d = \|p - x_0\| * \sin(\theta) = \frac{\|v\| * \|p - x_0\| * \sin(\theta)}{\|v\|} = \frac{\|v \times (p - x_0)\|}{\|v\|}$$

**Theorem 1.4.4: Distance: Parallel Planes**

The distance between parallel planes  $P_1: a(x - x_1) + b(y - y_1) + c(z - z_1) = 0$

and  $P_2: a(x - x_2) + b(y - y_2) + c(z - z_2) = 0$ :

$$d = \frac{|(a,b,c) \cdot (x_2 - x_1, y_2 - y_1, z_2 - z_1)|}{\sqrt{a^2 + b^2 + c^2}}$$

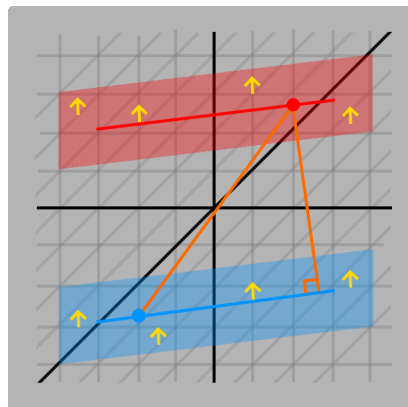
**Proof**

Planes  $P_1$  and  $P_2$  are parallel since they both have the normal vector  $n = (a, b, c)$ . Since  $(x_1, y_1, z_1)$  is a point on  $P_1$  and  $(x_2, y_2, z_2)$  is a point on  $P_2$ , then  $(x_2, y_2, z_2) - (x_1, y_1, z_1)$  is a vector from  $P_1$  to  $P_2$ .

Then the distance is the norm of the orthogonal component of  $(x_2, y_2, z_2) - (x_1, y_1, z_1)$  to  $P_1, P_2$ . Since normal vector  $n$  is orthogonal to both planes, then the orthogonal component of  $(x_2, y_2, z_2) - (x_1, y_1, z_1)$  and  $n$  are parallel. Thus, by **theorem 1.2.4**:

$$d = \|\text{proj}_n[(x_2, y_2, z_2) - (x_1, y_1, z_1)]\| = \left\| \frac{[(x_2, y_2, z_2) - (x_1, y_1, z_1)] \cdot (a, b, c)}{\|(a, b, c)\|^2} (a, b, c) \right\|$$

$$d = \frac{|(x_2 - x_1, y_2 - y_1, z_2 - z_1) \cdot (a, b, c)|}{\|(a, b, c)\|}$$



**Theorem 1.4.5: Distance: Skew Lines**

Lines  $L_1, L_2 \in \mathbb{R}^3$  are skewed if they are neither parallel or intersecting.

Let  $L_1(t) = tv_1 + x_1$  and  $L_2(t) = tv_2 + x_2$ . The distance between  $L_1$  and  $L_2$ :

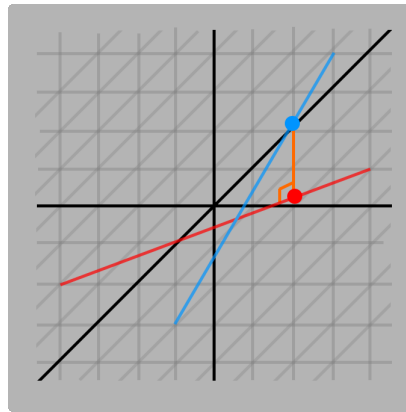
$$d = \frac{|(v_2 \times v_1) \cdot (x_2 - x_1)|}{\|v_2 \times v_1\|}$$

**Proof**

Let  $L_1, L_2$  be in two parallel planes. Note the distance between  $L_1$  and  $L_2$  is the distance between the two planes.

Since  $v_2 \times v_1$  is orthogonal to  $v_1, v_2$  and  $v_1, v_2$  are vectors parallel to each plane, then  $v_2 \times v_1$  is orthogonal to each plane and thus, a normal vector. By **theorem 1.4.4**:

$$d = \frac{|(v_2 \times v_1) \cdot (x_2 - x_1)|}{\|v_2 \times v_1\|}$$

**1.5 Matrices****Definition 1.5.1: Matrix**

A **m by n matrix**  $M_{m \times n}(\mathbb{R})$ :

$$M = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{bmatrix} \quad \text{where each } a_{ij} \in \mathbb{R}$$

A **row vector** is a 1 by n matrix:

$$\begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \end{bmatrix}$$

A **column vector** is a m by 1 matrix:

$$\begin{bmatrix} a_{11} \\ a_{21} \\ \vdots \\ a_{m1} \end{bmatrix}$$

A **zero matrix**  $0 \in M_{m \times n}(\mathbb{R})$ :

$$0 = \begin{bmatrix} 0_{11} & 0_{12} & \dots & 0_{1n} \\ 0_{21} & 0_{22} & \dots & 0_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ 0_{m1} & 0_{m2} & \dots & 0_{mn} \end{bmatrix}$$

**Theorem 1.5.2: Matrix Operations**

(a) Addition

For  $A, B \in M_{m \times n}(\mathbb{R})$ , then  $A+B \in M_{m \times n}(\mathbb{R})$  where each  $a_{ij}, b_{ij}$ :

$$\begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{bmatrix} + \begin{bmatrix} b_{11} & b_{12} & \dots & b_{1n} \\ b_{21} & b_{22} & \dots & b_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ b_{m1} & b_{m2} & \dots & b_{mn} \end{bmatrix} = \begin{bmatrix} a_{11} + b_{11} & a_{12} + b_{12} & \dots & a_{1n} + b_{1n} \\ a_{21} + b_{21} & a_{22} + b_{22} & \dots & a_{2n} + b_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} + b_{m1} & a_{m2} + b_{m2} & \dots & a_{mn} + b_{mn} \end{bmatrix}$$

(b) Scalar Multiplication

For  $A \in M_{m \times n}(\mathbb{R})$ , then  $cA \in M_{m \times n}(\mathbb{R})$  where each  $a_{ij}, b_{ij}$ :

$$c \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{bmatrix} = \begin{bmatrix} ca_{11} & ca_{12} & \dots & ca_{1n} \\ ca_{21} & ca_{22} & \dots & ca_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ ca_{m1} & ca_{m2} & \dots & ca_{mn} \end{bmatrix}$$

(c) Multiplication

For  $A \in M_{m \times n}(\mathbb{R})$ ,  $B \in M_{n \times k}(\mathbb{R})$ , then  $AB \in M_{m \times k}(\mathbb{R})$  where each  $a_{ij}, b_{ij}$ :

$$\begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{bmatrix} \begin{bmatrix} b_{11} & b_{12} & \dots & b_{1k} \\ b_{21} & b_{22} & \dots & b_{2k} \\ \vdots & \vdots & \ddots & \vdots \\ b_{n1} & b_{n2} & \dots & b_{nk} \end{bmatrix} = \begin{bmatrix} \sum_{i=1}^n a_{1i}b_{i1} & \sum_{i=1}^n a_{1i}b_{i2} & \dots & \sum_{i=1}^n a_{1i}b_{ik} \\ \sum_{i=1}^n a_{2i}b_{i1} & \sum_{i=1}^n a_{2i}b_{i2} & \dots & \sum_{i=1}^n a_{2i}b_{ik} \\ \vdots & \vdots & \ddots & \vdots \\ \sum_{i=1}^n a_{mi}b_{i1} & \sum_{i=1}^n a_{mi}b_{i2} & \dots & \sum_{i=1}^n a_{mi}b_{ik} \end{bmatrix}$$

**Theorem 1.5.3: Properties of Matrix Operations**(a)  $A+B = B+A$ Proof

$$[A+B]_{ij} = a_{ij} + b_{ij} = b_{ij} + a_{ij} = [B+A]_{ij}$$

(b)  $A+(B+C) = (A+B)+C$ Proof

$$[A+(B+C)]_{ij} = a_{ij} + (b_{ij} + c_{ij}) = (a_{ij} + b_{ij}) + c_{ij} = [(A+B)+C]_{ij}$$

(c)  $A+0 = A$ Proof

$$[A+0]_{ij} = a_{ij} + 0_{ij} = a_{ij} = [A]_{ij}$$

(d)  $(c+k)A = cA + kA$ Proof

$$[(c+k)A]_{ij} = (c+k)a_{ij} = ca_{ij} + ka_{ij} = [cA]_{ij} + [kA]_{ij} = [cA+kA]_{ij}$$

(e)  $c(A+B) = cA + cB$ Proof

$$[c(A+B)]_{ij} = c(a_{ij} + b_{ij}) = ca_{ij} + cb_{ij} = [cA]_{ij} + [cB]_{ij} = [cA+cB]_{ij}$$

(f)  $c(kA) = (ck)A = k(cA)$ Proof

$$[c(kA)]_{ij} = c(ka_{ij}) = (ck)a_{ij} = [(ck)A]_{ij} = k(ca_{ij}) = [k(cA)]_{ij}$$

(g)  $A(BC) = (AB)C$

**Proof**

$$\begin{aligned} &\text{Let } A \in M_{m \times n}(\mathbb{R}), B \in M_{n \times k}(\mathbb{R}), \text{ and } C \in M_{k \times p}(\mathbb{R}). \\ &\text{For } u \in \{1, \dots, n\} \text{ and } v \in \{1, \dots, p\}, \text{ then } [BC]_{uv} = \sum_{s=1}^k b_{us}c_{sv}. \\ &\text{Thus, for } i \in \{1, \dots, m\} \text{ and } j \in \{1, \dots, p\}: \\ &[A(BC)]_{ij} = \sum_{t=1}^n a_{it}[BC]_{tj} = \sum_{t=1}^n [a_{it} \sum_{s=1}^k b_{ts}c_{sj}] = \sum_{t=1}^n \sum_{s=1}^k a_{it}b_{ts}c_{sj} \\ &= \sum_{s=1}^k \sum_{t=1}^n a_{it}b_{ts}c_{sj} = \sum_{s=1}^k [\sum_{t=1}^n a_{it}b_{ts}]c_{sj} = \sum_{s=1}^k [AB]_{is}c_{sj} = [(AB)C]_{ij} \end{aligned}$$

(h)  $c(AB) = (cA)B = A(cB)$

**Proof**

$$\begin{aligned} &\text{Let } A \in M_{m \times n}(\mathbb{R}) \text{ and } B \in M_{n \times k}(\mathbb{R}). \text{ For } i \in \{1, \dots, m\} \text{ and } j \in \{1, \dots, k\}: \\ &[c(AB)]_{ij} = c \sum_{t=1}^n a_{it}b_{tj} = \sum_{t=1}^n ca_{it}b_{tj} = \sum_{t=1}^n (ca_{it})b_{tj} = [(cA)B]_{ij} \\ &[c(AB)]_{ij} = c \sum_{t=1}^n a_{it}b_{tj} = \sum_{t=1}^n a_{it}cb_{tj} = \sum_{t=1}^n a_{it}(cb_{tj}) = [A(cB)]_{ij} \end{aligned}$$

(i)  $A(B+C) = AB + AC$

**Proof**

$$\begin{aligned} &\text{Let } A \in M_{m \times n}(\mathbb{R}) \text{ and } B, C \in M_{n \times k}(\mathbb{R}). \text{ For } i \in \{1, \dots, m\} \text{ and } j \in \{1, \dots, k\}: \\ &[A(B+C)]_{ij} = \sum_{t=1}^n a_{it}[B+C]_{tj} = \sum_{t=1}^n a_{it}(b_{tj} + c_{tj}) = \sum_{t=1}^n a_{it}b_{tj} + a_{it}c_{tj} \\ &= \sum_{t=1}^n a_{it}b_{tj} + \sum_{t=1}^n a_{it}c_{tj} = [AB]_{ij} + [AC]_{ij} = [AB + AC]_{ij} \end{aligned}$$

(j)  $(A+B)C = AC + BC$

**Proof**

$$\begin{aligned} &\text{Let } A, B \in M_{m \times n}(\mathbb{R}) \text{ and } C \in M_{n \times k}(\mathbb{R}). \\ &\text{For } i \in \{1, \dots, m\} \text{ and } j \in \{1, \dots, k\}, \text{ the } ij\text{-th entry for } (A+B)C: \\ &[(A+B)C]_{ij} = \sum_{t=1}^n [A+B]_{it}c_{tj} = \sum_{t=1}^n (a_{it} + b_{it})c_{tj} = \sum_{t=1}^n a_{it}c_{tj} + b_{it}c_{tj} \\ &= \sum_{t=1}^n a_{it}c_{tj} + \sum_{t=1}^n b_{it}c_{tj} = [AC]_{ij} + [BC]_{ij} = [AC + BC]_{ij} \end{aligned}$$

**Definition 1.5.4: Transpose**For matrix  $A \in M_{m \times n}(\mathbb{R})$ :

$$A = \begin{bmatrix} a_{11} & a_{12} & a_{13} & \dots & a_{1n} \\ a_{21} & a_{22} & a_{23} & \dots & a_{2n} \\ a_{31} & a_{32} & a_{33} & \dots & a_{3n} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & a_{m3} & \dots & a_{mn} \end{bmatrix}$$

then the **transpose**,  $A^T \in M_{n \times m}(\mathbb{R})$ :

$$A^T = \begin{bmatrix} a_{11} & a_{21} & a_{31} & \dots & a_{m1} \\ a_{12} & a_{22} & a_{32} & \dots & a_{m2} \\ a_{13} & a_{23} & a_{33} & \dots & a_{m3} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ a_{1n} & a_{2n} & a_{3n} & \dots & a_{mn} \end{bmatrix}$$

**Theorem 1.5.5: Properties of the Transpose**

(a)  $(A^T)^T = A$

**Proof**

$$[(A^T)^T]_{ij} = [A^T]_{ji} = [A]_{ij}$$

(b)  $(AB)^T = B^T A^T$

**Proof**

$$\text{Let } A \in M_{m \times n}(\mathbb{R}) \text{ and } B \in M_{n \times k}(\mathbb{R}). \text{ For } i = \{1, \dots, k\} \text{ and } j = \{1, \dots, m\}: \\ [(AB)^T]_{ij} = [AB]_{ji} = \sum_{t=1}^n a_{jt} b_{ti} = \sum_{t=1}^n b_{ti} a_{jt} = \sum_{t=1}^n b_{it}^T a_{tj}^T = [B^T A^T]_{ij}$$

(c)  $x \cdot y = x^T y$

**Proof**

$$x \cdot y = x_1 y_1 + x_2 y_2 + \dots + x_n y_n = [x_1 \ x_2 \ \dots \ x_n] y = x^T y$$

**Definition 1.5.6: Determinant**

For  $A \in M_{n \times n}(\mathbb{R})$ , let  $\text{prod}(A) = a_{1,j_1} * a_{2,j_2} * \dots * a_{n,j_n}$  such that for any two  $a_{k,j_k}, a_{p,j_p}$  where  $k < p$ , then  $j_k \neq j_p$ . Let  $\text{prod}(A)$  be unique in the sense that no two  $\text{prod}(A)$  have exactly the same  $\{a_{1,j_1}, a_{2,j_2}, \dots, a_{n,j_n}\}$ .

Also, for any two such  $a_{k,j_k}, a_{p,j_p}$ , let an inversion be 1 if  $j_k < j_p$  and 0 if  $j_k > j_p$ . Then for any  $\text{prod}(A)$ , associate a  $\text{sign}(A) = (-1)^{\text{total number of inversions in prod}(A)}$ .

Then the **determinant** of  $A$ :

$$\det(A) = \sum_{\text{all prod}(A)} \text{prod}(A) * \text{sign}(A)$$

**Example**

$$\text{Let } A = \begin{bmatrix} 1 & 2 & 3 \\ -1 & 1 & 1 \\ 5 & -2 & 3 \end{bmatrix}. \text{ Find the } \det(A).$$

$$\det(A) = (1*1*3)(-1)^0 + (1*-2*1)(-1)^1 + (-1*2*3)(-1)^1 + (-1*-2*3)(-1)^2 \\ + (5*2*1)(-1)^2 + (5*1*3)(-1)^3 = 12$$

**Theorem 1.5.7: Cofactor Expansion**

Let  $A \in M_{n \times n}(\mathbb{R})$ . Let  $A_{ij}$  be  $A$ , but the  $i$ -th row and  $j$ -th column removed.

Then for a fixed  $i \in \{1, \dots, n\}$ :

$$\det(A) = (-1)^{i+1}a_{i1}\det(A_{i1}) + (-1)^{i+2}a_{i2}\det(A_{i2}) + \dots + (-1)^{i+n}a_{in}\det(A_{in})$$

Or for a fixed  $j \in \{1, \dots, n\}$ :

$$\det(A) = (-1)^{1+j}a_{1j}\det(A_{1j}) + (-1)^{2+j}a_{2j}\det(A_{2j}) + \dots + (-1)^{n+j}a_{nj}\det(A_{nj})$$

**Proof**

For any  $n$  by  $n$  matrix  $A$ , each  $\text{prod}(A)$  must contain  $n$   $a_{ij}$  where each  $a_{ij}$ 's  $i, j$  is different from another  $a_{ij}$ 's  $i, j$ . Thus, each  $\text{prod}(A)$  must contain only one  $a_{ij}$  in each row and column.

There are  $n$  possible  $a_{ij}$  choices in the first column and by choosing any such one, then that row is eliminated for choice in the following columns. Thus, there are  $n-1$  possible  $a_{ij}$  choices in the second column and by choosing any such one, then that row is also eliminated for choice in the following columns. Repeating the pattern, then there are  $n \cdot (n-1) \cdot (n-2) \cdot \dots \cdot 1 = n!$  total unique  $\text{prod}(A)$  combinations. In the cofactor expansion, let choose a fixed  $i$ . The case for a fixed  $j$  is analogous. For a fixed  $i$ , the cofactor expansion iterates through each of the  $n$  columns in row  $i$  so there are  $n$  unique  $a_{ij}$ . For each  $a_{ij}$ , the  $A_{ij}$  has the  $i$ -th row and  $j$ -th column removed so  $A_{ij}$  is a  $(n-1)$  by  $(n-1)$  matrix and thus, there are  $(n-1)!$  unique  $\text{prod}(A_{ij})$  combinations as proved earlier. Since each  $A_{ij}$  removes a different  $j$ -th column, then each  $\text{prod}(A_{ij})$  from different columns are unique. Thus, the  $n$  unique  $a_{ij}$  has  $(n-1)!$  unique  $\text{prod}(A_{ij})$  combinations so there are  $n \cdot (n-1)! = n!$  unique  $\text{prod}(A)$  combinations. Thus, the  $\text{prod}(A)$  combinations in the cofactor expansion must be equivalent to the  $\text{prod}(A)$  combinations in the original determinant.

For the fixed  $i$ , let fixed  $j \in \{1, \dots, n\}$ :

$$\begin{bmatrix} a_{1,1} & a_{1,2} & a_{1,3} & \dots & a_{1,j-1} & a_{1,j} & a_{1,j+1} & \dots & a_{1,n} \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots & \vdots & \ddots & \vdots \\ a_{i-1,1} & a_{i-1,2} & a_{i-1,3} & \dots & a_{i-1,j-1} & a_{i-1,j} & a_{i-1,j+1} & \dots & a_{i-1,n} \\ a_{i,1} & a_{i,2} & a_{i,3} & \dots & a_{i,j-1} & a_{i,j} & a_{i,j+1} & \dots & a_{i,n} \\ a_{i+1,1} & a_{i+1,2} & a_{i+1,3} & \dots & a_{i+1,j-1} & a_{i+1,j} & a_{i+1,j+1} & \dots & a_{i+1,n} \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots & \vdots & \ddots & \vdots \\ a_{n,1} & a_{n,2} & a_{n,3} & \dots & a_{n,j-1} & a_{n,j} & a_{n,j+1} & \dots & a_{n,n} \end{bmatrix}$$

In the original determinant, each  $\text{prod}(A)$  associates  $\text{sign}(A) = (-1)^{\# \text{ inversions in } \text{prod}(A)}$ .

As proven earlier, each  $\text{prod}(A)$  is expressed in the cofactor expansion. So for any  $\text{prod}(A)$  that contains  $a_{ij}$  with the fixed  $i, j$ , then from the  $a_{ij}\det(A_{ij})$  in the cofactor expansion, the  $\det(A_{ij})$  consists of the other  $a_{ij}$  in the  $\text{prod}(A)$  since none of the other  $a_{ij}$  can exist in row  $i$  or column  $j$  by definition of the determinant and thus,  $\det(A_{ij})$  must account for all the inversions exclusively between the other  $a_{ij}$ . To account for the inversions between the other  $a_{ij}$  and the fixed  $a_{ij}$ , refer to the matrix above. The only  $a_{ij}$  which contributes an inversion with the fixed  $a_{ij}$  must be in the lower left and upper right of the matrix by definition of the determinant. Let  $A = \#a_{ij}$  in upper left,  $B = \#a_{ij}$  in upper right,  $C = \#a_{ij}$  in lower left, and  $D = \#a_{ij}$  in lower right.

Since each  $\text{prod}(A)$  must have a  $a_{ij}$  in each row and column, then:

$$A+B = i-1 \quad A+C = j-1 \quad \Rightarrow \quad B+C = i+j-2-2A$$

Thus,  $\text{sign}(A) = (-1)^{B+C} = (-1)^{i+j-2-2A} = (-1)^{i+j}(-1)^{-2}(-1)^{-2A} = (-1)^{i+j}$  which is the coefficient in the cofactor expansion and thus, the cofactor expansion is calculated in the same way as the original determinant and thus, have the same value.



**Example**

Let  $A = \begin{bmatrix} 1 & 2 & 3 \\ -1 & 1 & 1 \\ 5 & -2 & 3 \end{bmatrix}$ . Find the  $\det(A)$ .

Using row 1:

$$\det(A) = 1 \begin{vmatrix} 1 & 1 \\ -2 & 3 \end{vmatrix} - 2 \begin{vmatrix} -1 & 1 \\ 5 & 3 \end{vmatrix} + 3 \begin{vmatrix} -1 & 1 \\ 5 & -2 \end{vmatrix} = 1*5 - 2*-8 + 3*-3 = 12$$

Using column 2:

$$\det(A) = -2 \begin{vmatrix} -1 & 1 \\ 5 & 3 \end{vmatrix} + 1 \begin{vmatrix} 1 & 3 \\ 5 & 3 \end{vmatrix} - (-2) \begin{vmatrix} 1 & 3 \\ -1 & 1 \end{vmatrix} = -2*-8 + 1*-12 - (-2)*4 = 12$$

**1.6 Different Coordinate Systems****Definition 1.6.1: Polar Coordinates**

Thus far, all vectors has been in the Cartesian (i.e. rectangular (x,y)) System. However, vectors can also be expressed in the Polar (i.e. circular) System.

For any point (x,y), a right triangle can be drawn by adding a perpendicular line from the x-axis to (x,y). Thus:

$$r = \sqrt{x^2 + y^2} \quad x = r \cos(\theta) \quad y = r \sin(\theta)$$

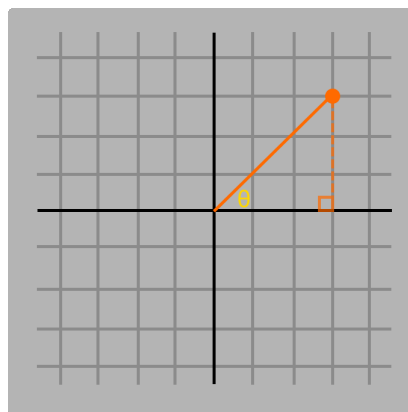
Thus, the **polar coordinates** can express points as  $(r, \theta)$ .

To convert from polar to rectangular:

$$x = r \cos(\theta) \quad y = r \sin(\theta)$$

To convert from rectangular to polar:

$$r^2 = x^2 + y^2 \quad \tan(\theta) = \frac{y}{x}$$

**Example**

Convert  $(x + 1)^2 + (y - 2)^2 = 9$  into polar coordinates.

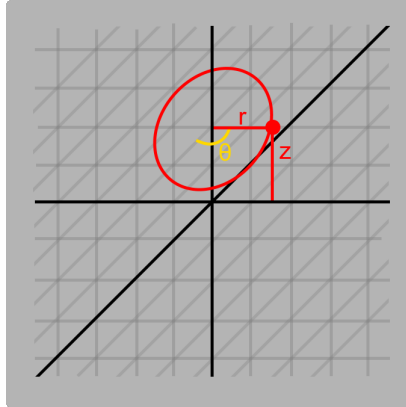
$$\begin{aligned} x^2 + 2x + 1 + y^2 - 4y + 4 &= 9 \\ r^2 + 2r \cos(\theta) - 4r \sin(\theta) &= 4 \end{aligned}$$

**Definition 1.6.2: Cylindrical Coordinates**

While polar coordinates are the circular equivalent to  $\mathbb{R}^2$ , cylindrical coordinates are the circular equivalent to  $\mathbb{R}^3$ .

**Cylindrical coordinates** are expressed as  $(r, \theta, z)$  where:

$$x = r \cos(\theta) \quad y = r \sin(\theta) \quad z = z$$

**Example**

Convert  $z = 5r$  to rectangular coordinates.

$$z^2 = 25r^2 \quad \Rightarrow \quad z^2 = 25(x^2 + y^2) \quad \Rightarrow \quad z = 5\sqrt{x^2 + y^2}$$

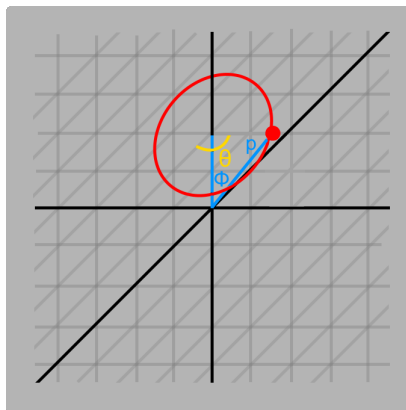
**Definition 1.6.3: Spherical Coordinates**

Another way to express coordinates in  $\mathbb{R}^3$  is spherical coordinates.

**Spherical coordinates** are expressed as  $(p, \theta, \phi)$  where:

$$x = p \sin(\phi) \cos(\theta) \quad y = p \sin(\phi) \sin(\theta) \quad z = p \cos(\phi)$$

where  $p = \sqrt{x^2 + y^2 + z^2} \geq 0$ ,  $\theta \in [0, 2\pi]$ , and  $\phi \in [0, \pi]$

**Example**

Convert  $z = 5r$  to spherical coordinates.

$$z = 5r \quad \Rightarrow \quad \frac{r}{z} = \frac{1}{5} \quad \Rightarrow \quad \tan(\phi) = \frac{1}{5}$$

## 2 Differentiation

### 2.1 Limits & Continuity

#### Definition 2.1.1: Limit

For  $f: X \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$ , let  $a \in X$ .

If for every  $\epsilon > 0$ , there is a  $\delta > 0$  such that for all  $x \in X$  where  $\|x - a\| < \delta$ :

$$\|f(x) - L\| < \epsilon$$

Then the **limit** of  $f(x)$  as  $x$  approaches  $a$  is  $\lim_{x \rightarrow a} f(x) = L$ .

#### Example

Let  $f(x,y) = 2x^2 + xy$ . Find  $f(x,y)$  as  $(x,y) \rightarrow (-1,1)$ .

$$\begin{aligned} L &= f(-1,1) = 1. \text{ Let } \sqrt{(x+1)^2 + (y-1)^2} < \delta \text{ so } |x+1| < \delta \text{ and } |y-1| < \delta. \text{ Thus:} \\ |f(x,y) - L| &= |2x^2 + xy - 1| = |2x^2 - 2 + xy + 1| \\ &= |2(x+1)(x-1) + (x+1)(y+1) - (x+1+y-1)| \\ &\leq 2|x+1| * |x-1| + |x+1| * |y+1| + |x+1| + |y-1| \\ &< 2\delta(\delta+2) + \delta(\delta+2) + 2\delta = 3\delta^2 + 8\delta \\ \text{Since } \min(3\delta^2 + 8\delta) &= \frac{-16}{3} < 0, \text{ then for any } \epsilon > 0, \text{ there is a } \delta \text{ where } 3\delta^2 + 8\delta < \epsilon. \\ \text{Thus, } |f(x,y) - L| &< 3\delta^2 + 8\delta < \epsilon. \end{aligned}$$

#### Theorem 2.1.2: Limits are Unique

If  $\lim_{x \rightarrow a} f(x) = L_1$  and  $\lim_{x \rightarrow a} f(x) = L_2$ , then  $L_1 = L_2$ .

#### Proof

$$\begin{aligned} \text{Since } \lim_{x \rightarrow a} f(x) &= L_1, \text{ there is a } \delta_1 \text{ where for } \|x - a\| < \delta_1, \text{ then } \|f(x) - L_1\| < \frac{\epsilon}{2}. \\ \text{Since } \lim_{x \rightarrow a} f(x) &= L_2, \text{ there is a } \delta_2 \text{ where for } \|x - a\| < \delta_2, \text{ then } \|f(x) - L_2\| < \frac{\epsilon}{2}. \\ \text{Let } \delta &= \min(\delta_1, \delta_2). \text{ Then for } \|x - a\| < \delta: \\ \|L_1 - L_2\| &\leq \|L_1 - f(x)\| + \|f(x) - L_2\| < \frac{\epsilon}{2} + \frac{\epsilon}{2} = \epsilon \end{aligned}$$

#### Theorem 2.1.3: Properties of the Limit

(a) For  $f, g: \mathbb{R}^n \rightarrow \mathbb{R}^m$ , if  $\lim_{x \rightarrow a} f(x) = A$  and  $\lim_{x \rightarrow a} g(x) = B$ , then:

$$\lim_{x \rightarrow a} (f+g)(x) = A+B$$

#### Proof

$$\begin{aligned} \text{Since } \lim_{x \rightarrow a} f(x) &= A, \text{ there is a } \delta_1 \text{ where for } \|x - a\| < \delta_1, \text{ then:} \\ \|f(x) - A\| &< \frac{\epsilon}{2} \\ \text{Since } \lim_{x \rightarrow a} g(x) &= B, \text{ there is a } \delta_2 \text{ where for } \|x - a\| < \delta_2, \text{ then:} \\ \|g(x) - B\| &< \frac{\epsilon}{2} \\ \text{Let } \delta &= \min(\delta_1, \delta_2). \text{ Then for } \|x - a\| < \delta: \\ \|(f+g)(x) - (A+B)\| &= \|f(x) + g(x) - A - B\| \\ &\leq \|f(x) - A\| + \|g(x) - B\| < \frac{\epsilon}{2} + \frac{\epsilon}{2} = \epsilon \end{aligned}$$

(b) For  $f: \mathbb{R}^n \rightarrow \mathbb{R}^m$ , if  $\lim_{x \rightarrow a} f(x) = A$  and scalar  $c \in \mathbb{R}$ , then:

$$\lim_{x \rightarrow a} cf(x) = cA$$

#### Proof

$$\begin{aligned} \text{Since } \lim_{x \rightarrow a} f(x) &= A, \text{ there is a } \delta \text{ where for } \|x - a\| < \delta, \text{ then:} \\ \|f(x) - A\| &< \frac{\epsilon}{c} \\ \text{Then, } \|cf(x) - cA\| &= c\|f(x) - A\| < c\frac{\epsilon}{c} = \epsilon. \end{aligned}$$

- (c) For  $f, g: \mathbb{R}^n \rightarrow \mathbb{R}$ , if  $\lim_{x \rightarrow a} f(x) = A$  and  $\lim_{x \rightarrow a} g(x) = B$ , then:  
 $\lim_{x \rightarrow a} (fg)(x) = AB$

**Proof**

Note  $4fg = (f + g)^2 - (f - g)^2$ .

By part (a), there is a  $\delta$  where for  $\|x - a\| < \delta$ :

$$|(f + g)(x) - (A + B)| < \epsilon$$

Then as  $x \rightarrow a$ :

$$\begin{aligned} |[(f + g)(x)]^2 - [A + B]^2| &= |[(f + g)(x) - (A + B)][(f + g)(x) + (A + B)]| \\ &= |(f + g)(x) - (A + B)| * |(f + g)(x) + (A + B)| = \epsilon(2(A + B)) \end{aligned}$$

Thus,  $\lim_{x \rightarrow a} (f + g)^2(x) = (A + B)^2$ .

The proof for  $\lim_{x \rightarrow a} (f - g)^2(x) = (A - B)^2$  is analogous. Thus:

$$\begin{aligned} \lim_{x \rightarrow a} (fg)(x) &= \lim_{x \rightarrow a} \frac{1}{4}[(f + g)^2(x) - (f - g)^2(x)] \\ &= \frac{1}{4}[(A + B)^2 - (A - B)^2] = \frac{1}{4}4AB = AB \end{aligned}$$

- (d) For  $f, g: \mathbb{R}^n \rightarrow \mathbb{R}$ , if  $\lim_{x \rightarrow a} f(x) = A$  and  $\lim_{x \rightarrow a} g(x) = B \neq 0$ , then:  
 $\lim_{x \rightarrow a} \left(\frac{f}{g}\right)(x) = \frac{A}{B}$

**Proof**

Since  $\lim_{x \rightarrow a} g(x) = B$ , then there is a  $\delta$  where for  $\|x - a\| < \delta$ :

$$|g(x) - B| < \epsilon$$

Thus, as  $x \rightarrow a$ :

$$\left|\frac{1}{g(x)} - \frac{1}{B}\right| = \left|\frac{B - g(x)}{Bg(x)}\right| = |B - g(x)| * \left|\frac{1}{Bg(x)}\right| < \epsilon \frac{1}{B^2}$$

Thus,  $\lim_{x \rightarrow a} \frac{1}{g(x)} = \frac{1}{B}$ . By part (c), then  $\lim_{x \rightarrow a} \left(\frac{f}{g}\right)(x) = \frac{A}{B}$ .

#### Theorem 2.1.4: Components of Limits

For  $f: X \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$ , let  $f(x) = (f_1(x), \dots, f_m(x))$ . Then for  $i = \{1, \dots, m\}$ :  
 $\lim_{x \rightarrow a} f(x) = L = (L_1, \dots, L_m)$  if and only if each  $\lim_{x \rightarrow a} f_i(x) = L_i$

**Proof**

If  $\lim_{x \rightarrow a} f(x) = L = (L_1, \dots, L_m)$ , then there is a  $\delta$  such that for  $\|x - a\| < \delta$ :

$$\|f(x) - L\| < \epsilon$$

$$\|(f_1(x), \dots, f_m(x)) - (L_1, \dots, L_m)\| = \sqrt{(f_1(x) - L_1)^2 + \dots + (f_m(x) - L_m)^2} < \epsilon$$

Thus, each  $|f_i(x) - L_i| < \epsilon$  for  $\|x - a\| < \delta$  so  $\lim_{x \rightarrow a} f_i(x) = L_i$ .

If each  $\lim_{x \rightarrow a} f_i(x) = L_i$ , then there are  $\delta_i$  such that for  $\|x - a\| < \delta_i$ :

$$|f_i(x) - L_i| < \frac{\epsilon}{\sqrt{m}}$$

Let  $\delta = \min(\delta_1, \dots, \delta_m)$ . Then for  $\|x - a\| < \delta$ :

$$\begin{aligned} \|f(x) - L\| &= \|(f_1(x), \dots, f_m(x)) - (L_1, \dots, L_m)\| \\ &= \sqrt{(f_1(x) - L_1)^2 + \dots + (f_m(x) - L_m)^2} < \sqrt{\sum_{i=1}^m \left(\frac{\epsilon}{\sqrt{m}}\right)^2} = \sqrt{\epsilon^2} = \epsilon \end{aligned}$$

#### Definition 2.1.5: Continuity

For  $f: X \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$ , let  $a \in X$ .

Then  $f$  is **continuous** at  $a$  if  $\lim_{x \rightarrow a} f(x) = f(a)$ .

If  $f$  is continuous at every  $x \in X$ , then  $f$  is continuous on  $X$ .

**Theorem 2.1.6: Properties of Continuity**

- (a) If
- $f, g: X \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$
- are continuous at
- $a \in X$
- , then
- $f+g$
- is continuous at
- $a$

**Proof**

Since  $\lim_{x \rightarrow a} f(x) = f(a)$  and  $\lim_{x \rightarrow a} g(x) = g(a)$ , by **theorem 2.1.3(a)**, then  $A = f(a)$  and  $B = g(a)$ . Thus,  $\lim_{x \rightarrow a} (f+g)(x) = f(a)+g(a)$ .

- (b) If
- $f: X \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$
- is continuous at
- $a \in X$
- and scalar
- $c \in \mathbb{R}$
- , then
- $cf$
- is continuous at
- $a$

**Proof**

Since  $\lim_{x \rightarrow a} f(x) = f(a)$ , by **theorem 2.1.3(b)**, then  $A = f(a)$ .  
Thus,  $\lim_{x \rightarrow a} cf(x) = cf(a)$ .

- (c) If
- $f, g: X \subset \mathbb{R}^n \rightarrow \mathbb{R}$
- are continuous at
- $a \in X$
- , then
- $fg$
- is continuous at
- $a$

**Proof**

Since  $\lim_{x \rightarrow a} f(x) = f(a)$  and  $\lim_{x \rightarrow a} g(x) = g(a)$ , by **theorem 2.1.3(c)**, then  $A = f(a)$  and  $B = g(a)$ . Thus,  $\lim_{x \rightarrow a} (fg)(x) = f(a)g(a)$ .

- (d) If
- $f, g: X \subset \mathbb{R}^n \rightarrow \mathbb{R}$
- are continuous at
- $a \in X$
- where
- $g(x) \neq 0$
- , then
- $\frac{f}{g}$
- is continuous at
- $a$

**Proof**

Since  $\lim_{x \rightarrow a} f(x) = f(a)$  and  $\lim_{x \rightarrow a} g(x) = g(a)$ , by **theorem 2.1.3(d)**, then  $A = f(a)$  and  $B = g(a) \neq 0$ . Thus,  $\lim_{x \rightarrow a} \left(\frac{f}{g}\right)(x) = \frac{f(a)}{g(a)}$ .

**Theorem 2.1.7: Components of Continuity**

For  $f: X \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$ , let  $f(x) = (f_1(x), \dots, f_m(x))$ . Then for  $i = \{1, \dots, m\}$ :  
 $f$  is continuous at  $a \in X$  if and only if each  $f_i$  is continuous at  $a$

**Proof**

If  $f$  is continuous at  $a$ , then  $\lim_{x \rightarrow a} f(x) = f(a) = (f_1(a), \dots, f_m(a))$ . By **theorem 2.1.4**, then  $L = (f_1(a), \dots, f_m(a))$  so each  $L_i = f_i(a)$ . Thus, for each  $i = \{1, \dots, m\}$ :

$$\lim_{x \rightarrow a} f_i(x) = L_i = f_i(a)$$

If each  $f_i$  is continuous at  $a$ , then for  $i = \{1, \dots, m\}$ ,  $\lim_{x \rightarrow a} f_i(x) = f_i(a)$ . By **theorem 2.1.4**, then  $L = (f_1(a), \dots, f_m(a))$ . Thus:

$$\lim_{x \rightarrow a} f(x) = L = (f_1(a), \dots, f_m(a)) = f(a)$$

**Theorem 2.1.8: Composite of Continuous functions are Continuous**

If  $f: X \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$  and  $g: Y \subset \mathbb{R}^m \rightarrow \mathbb{R}^k$  are continuous where  $f(X) \subset Y$ ,  
then  $g \circ f: X \subset \mathbb{R}^n \rightarrow \mathbb{R}^k$  is continuous

**Proof**

For any  $a \in X$  and any  $\delta > 0$ , there is a  $\eta > 0$  such that for  $\|x - a\| < \eta$ :

$$\|f(x) - f(a)\| < \delta$$

Since  $f(X) \subset Y$ , then for any  $x \in X$ , then  $f(x) \in Y$ .

For any  $f(a) \in Y$  and any  $\epsilon > 0$ , there is a  $\delta > 0$  such that for  $\|y - f(a)\| < \delta$ :

$$\|g(y) - g(f(a))\| < \epsilon$$

Thus, for  $\|x - a\| < \eta$ , then  $\|g(f(x)) - g(f(a))\| < \epsilon$ .

## 2.2 Differentiability

### Definition 2.2.1: Partial Derivative

For  $f: X \subset \mathbb{R}^n \rightarrow \mathbb{R}$ , let  $x = (x_1, \dots, x_n) \in X$ .

For  $i = \{1, \dots, n\}$ , the **partial derivative** of  $f$  with respect to  $x_i$ :

$$D_i f = \frac{\partial f}{\partial x_i} = f_{x_i}(x) = \lim_{h \rightarrow 0} \frac{f(x_1, \dots, x_i + h, \dots, x_n) - f(x_1, \dots, x_n)}{h}$$

### Theorem 2.2.2: Tangent Plane

For  $f: X \subset \mathbb{R}^2 \rightarrow \mathbb{R}$ , let  $z = f(x, y)$ .

The **tangent plane** at  $(a, b, f(a, b))$  has an equation of the form:

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

### Proof

Since  $f_x(a, b)$  which is the change in  $z$  for every change in  $x$  is a tangent vector to  $f$  in direction of  $x$  at  $(a, b)$ , then  $(1, 0, f_x(a, b))$  is parallel to the tangent plane. Similarly,  $(0, 1, f_y(a, b))$  is parallel to the tangent plane.

Thus,  $(1, 0, f_x(a, b)) \times (0, 1, f_y(a, b)) = (-f_x(a, b), -f_y(a, b), 1)$  is orthogonal to the tangent plane. Thus, for any  $(x, y, z)$  in the plane:

$$\begin{aligned} (-f_x(a, b), -f_y(a, b), 1) \cdot [(x, y, z) - (a, b, f(a, b))] &= 0 \\ -f_x(a, b)(x - a) - f_y(a, b)(y - b) + z - f(a, b) &= 0 \\ z = f(a, b) + f_x(a, b)(x - a) + f_y(a, b)(y - b) \end{aligned}$$

### Definition 2.2.3: Differentiability in $\mathbb{R}^2 \rightarrow \mathbb{R}$

$f: X \subset \mathbb{R}^2 \rightarrow \mathbb{R}$  is **differentiable** at  $x \in X$  if there is an  $A \in M_{1 \times 2}(\mathbb{R})$  such that for  $h \in X$ :

$$\lim_{h \rightarrow 0} \frac{|f(x+h) - f(x) - Ah|}{\|h\|} = 0$$

Then, the **derivative** of  $f$  at  $x$  is  $Df(x) = A = \begin{bmatrix} \frac{\partial f}{\partial x}(x, y) & \frac{\partial f}{\partial y}(x, y) \end{bmatrix}$ .

If  $f$  is differentiable at every  $x \in X$ , then  $f$  is differentiable on  $X$ .

### Theorem 2.2.4: Continuous partials imply Differentiability

If  $f: X \subset \mathbb{R}^2 \rightarrow \mathbb{R}$  has continuous partial derivatives at  $(a, b)$ , then  $f$  is differentiable at  $(a, b)$ .

### Proof

Since  $f_x(x, y), f_y(x, y)$  is continuous at  $(a, b)$ , then for  $\epsilon > 0$ , there is a  $\delta > 0$  where for  $\|(x, y) - (a, b)\| < \delta$ :

$$|f_x(x, y) - f_x(a, b)| < \epsilon \quad |f_y(x, y) - f_y(a, b)| < \epsilon$$

Then for  $h = h_1 e_1 + h_2 e_2$ :

$$\begin{aligned} \lim_{h \rightarrow 0} \frac{|f(a+h_1, b+h_2) - f(a, b) - [f_x(a, b)h_1 + f_y(a, b)h_2]|}{\|h\|} \\ = \lim_{h \rightarrow 0} \frac{|f(a+h_1, b+h_2) - f(a+h_1, b) + f(a+h_1, b) - f(a, b) - [f_x(a, b)h_1 + f_y(a, b)h_2]|}{\|h\|} \end{aligned}$$

Since  $f_x(x, y), f_y(x, y)$  exist, then by the Mean Value Theorem, there are  $t_1 \in (0, h_1)$  and  $t_2 \in (0, h_2)$  such that:

$$\begin{aligned} f(a+h_1, b) - f(a, b) &= h_1 * f_x(a+t_1, b) \\ f(a+h_1, b+h_2) - f(a+h_1, b) &= h_2 * f_y(a+h_1, b+t_2) \end{aligned}$$

Thus, for  $\|h - (a, b)\| < \delta$ :

$$\begin{aligned} \lim_{h \rightarrow 0} \frac{|f(a+h_1, b+h_2) - f(a, b) - [f_x(a, b)h_1 + f_y(a, b)h_2]|}{\|h\|} \\ = \lim_{h \rightarrow 0} \frac{|h_2 * f_y(a+h_1, b+t_2) + h_1 * f_x(a+t_1, b) - [f_x(a, b)h_1 + f_y(a, b)h_2]|}{\|h\|} \\ = \lim_{h \rightarrow 0} \frac{|h_2 * [f_y(a+h_1, b+t_2) - f_y(a, b)] + h_1 * [f_x(a+t_1, b) - f_x(a, b)]|}{\|h\|} < \lim_{h \rightarrow 0} \frac{\|h\|\epsilon + \|h\|\epsilon}{\|h\|} = 2\epsilon \end{aligned}$$

**Theorem 2.2.5: Differentiability implies Continuity**

If  $f: X \subset \mathbb{R}^2 \rightarrow \mathbb{R}$  is differentiable at  $(a,b)$ , then  $f$  is continuous at  $(a,b)$

**Proof**

If  $f$  is differentiable at  $(a,b)$ , then  $\lim_{h \rightarrow 0} \frac{|f((a,b)+h) - f(a,b) - Ah|}{\|h\|} = 0$ .

Thus, as  $h \rightarrow 0$ , then  $A = \frac{f((a,b)+h) - f(a,b)}{\|h\|}$ . So:

$$f((a,b)+h) - f(a,b) = [f((a,b)+h) - f(a,b)] \frac{\|h\|}{\|h\|} = A\|h\| \rightarrow 0$$

Thus,  $f$  is continuous at  $(a,b)$ .

**2.3 Differentiability in Higher Dimensions****Definition 2.3.1: Differentiability in  $\mathbb{R}^n \rightarrow \mathbb{R}$** 

Differentiability can be extended for  $\mathbb{R}^n$ .

$f: X \subset \mathbb{R}^n \rightarrow \mathbb{R}$  is differentiable at  $x \in X$  if there is an  $A \in M_{1 \times n}(\mathbb{R})$  such that for  $h \in X$ :

$$\lim_{h \rightarrow 0} \frac{|f(x+h) - f(x) - Ah|}{\|h\|} = 0$$

Then, the derivative of  $f$  at  $x$  is  $Df(x) = A = \left[ \frac{\partial f}{\partial x_1}(x) \quad \frac{\partial f}{\partial x_2}(x) \quad \dots \quad \frac{\partial f}{\partial x_n}(x) \right]$ .

If  $f$  is differentiable at every  $x \in X$ , then  $f$  is differentiable on  $X$ .

The **gradient** of  $f$ :

$$\nabla f(x) = \left( \frac{\partial f}{\partial x_1}(x), \dots, \frac{\partial f}{\partial x_n}(x) \right) = [Df(x)]^T$$

**Definition 2.3.2: Differentiability in  $\mathbb{R}^n \rightarrow \mathbb{R}^m$** 

Differentiability can be extended into  $\mathbb{R}^m$ .

$f: X \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$  where  $f = (f_1, \dots, f_m)$  is differentiable at  $x \in X$  if there is an  $A \in M_{m \times n}(\mathbb{R})$  such that for  $h \in X$ :

$$\lim_{h \rightarrow 0} \frac{|f(x+h) - f(x) - Ah|}{\|h\|} = 0$$

Then, the derivative of  $f$  at  $x$  is  $Df(x) = A = \begin{bmatrix} \frac{\partial f_1}{\partial x_1}(x) & \frac{\partial f_1}{\partial x_2}(x) & \dots & \frac{\partial f_1}{\partial x_n}(x) \\ \frac{\partial f_2}{\partial x_1}(x) & \frac{\partial f_2}{\partial x_2}(x) & \dots & \frac{\partial f_2}{\partial x_n}(x) \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial f_m}{\partial x_1}(x) & \frac{\partial f_m}{\partial x_2}(x) & \dots & \frac{\partial f_m}{\partial x_n}(x) \end{bmatrix}$ .

If  $f$  is differentiable at every  $x \in X$ , then  $f$  is differentiable on  $X$ .

**Example**

Let  $f(x,y) = (5xy, ye^x, x^2 + y^2)$ . Find the  $Df(x,y)$ .

$$Df(x,y) = \begin{bmatrix} \frac{\partial}{\partial x} 5xy & \frac{\partial}{\partial y} 5xy \\ \frac{\partial}{\partial x} ye^x & \frac{\partial}{\partial y} ye^x \\ \frac{\partial}{\partial x} x^2 + y^2 & \frac{\partial}{\partial y} x^2 + y^2 \end{bmatrix} = \begin{bmatrix} 5y & 5x \\ ye^x & e^x \\ 2x & 2y \end{bmatrix}$$

**Theorem 2.3.3: Differentiability implies Continuity in Higher Dimensions**

If  $f: X \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$  is differentiable at  $a \in X$ , then  $f$  is continuous at  $a$

**Proof**

Analogous to **theorem 2.2.5**. Replace  $(a,b)$  with  $a = (a_1, \dots, a_n)$ .

**Theorem 2.3.4: Continuous partials imply differentiability in Higher Dimensions**

If  $f: X \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$  has continuous partial derivatives,  $\frac{\partial f_i}{\partial x_j}$ , at  $a \in X$  for  $j = \{1, \dots, n\}$  and  $i = \{1, \dots, m\}$ , then  $f$  is differentiable at  $a$

**Proof**

Analogous to **theorem 2.2.4**. Instead,  $h = h_1 e_1 + \dots + h_n e_n$  where:

$$\lim_{h \rightarrow 0} \frac{|f(x+h) - f(x) - Ah|}{\|h\|} = \lim_{h \rightarrow 0} \sum_{i=1}^m \frac{|f_i(x+h) - f_i(x) - [\sum_{j=1}^n \frac{\partial f_i}{\partial x_j}(x) h_j]|}{\|h\|}$$

and add each  $f_i(x + h_1 e_1 + \dots + h_k e_k)$  and apply Mean Value Theorem and continuity of partial derivatives analogously as performed in **theorem 2.2.4**.

**Theorem 2.3.5: Components of Differentiability**

$f: X \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$  where  $f = (f_1, \dots, f_m)$  is differentiable at  $a \in X$  if and only if each  $f_i$  is differentiable at  $a$  for  $i = \{1, \dots, m\}$

**Proof**

$$\text{Note } \lim_{h \rightarrow 0} \frac{|f(x+h) - f(x) - Ah|}{\|h\|} = \lim_{h \rightarrow 0} \sum_{i=1}^m \frac{|f_i(x+h) - f_i(x) - [\sum_{j=1}^n \frac{\partial f_i}{\partial x_j}(x) h_j]|}{\|h\|}.$$

If  $f$  is differentiable at  $a$ , then for any  $\epsilon > 0$ :

$$\lim_{h \rightarrow 0} \frac{|f(x+h) - f(x) - Ah|}{\|h\|} < \epsilon$$

So  $\lim_{h \rightarrow 0} \frac{|f_i(x+h) - f_i(x) - [\sum_{j=1}^n \frac{\partial f_i}{\partial x_j}(x) h_j]|}{\|h\|} < \epsilon$  for each  $i = \{1, \dots, m\}$  and thus, each  $f_i$  is differentiable at  $a$ .

If each  $\lim_{h \rightarrow 0} \frac{|f_i(x+h) - f_i(x) - [\sum_{j=1}^n \frac{\partial f_i}{\partial x_j}(x) h_j]|}{\|h\|} < \frac{\epsilon}{m}$  for  $i = \{1, \dots, m\}$ , then:

$$\lim_{h \rightarrow 0} \frac{|f(x+h) - f(x) - Ah|}{\|h\|} < \lim_{h \rightarrow 0} \sum_{i=1}^m \frac{\epsilon}{m} = \epsilon$$

Thus,  $f$  is differentiable at  $a$ .

**Theorem 2.3.6: Properties of Differentiability**

- (a) For  $f, g: X \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$ , if  $f, g$  are differentiable at  $a \in X$ , then:  
 $f+g$  is differentiable at  $a$  where  $D(f+g)(a) = Df(a) + Dg(a)$

**Proof**

Since  $f, g$  are differentiable at  $a \in X$ , by **theorem 2.3.5**, then for  $i = \{1, \dots, m\}$ :

$$\begin{aligned} D(f+g)_i(a) &= \begin{bmatrix} D_1(f_i + g_i)(a) & D_2(f_i + g_i)(a) & \dots & D_n(f_i + g_i)(a) \end{bmatrix} \\ &= \begin{bmatrix} D_1 f_i(a) & D_2 f_i(a) & \dots & D_n f_i(a) \end{bmatrix} + \begin{bmatrix} D_1 g_i(a) & D_2 g_i(a) & \dots & D_n g_i(a) \end{bmatrix} \\ &= Df_i(a) + Dg_i(a) \end{aligned}$$

- (b) For  $f: X \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$ , if  $f$  is differentiable at  $a \in X$  and scalar  $c \in \mathbb{R}$ , then:  
 $cf$  is differentiable at  $a$  where  $D(cf)(a) = cDf(a)$

**Proof**

Since  $f$  is differentiable at  $a \in X$ , by **theorem 2.3.5**, then for  $i = \{1, \dots, m\}$ :

$$\begin{aligned} D(cf)_i(a) &= \begin{bmatrix} D_1(cf_i)(a) & D_2(cf_i)(a) & \dots & D_n(cf_i)(a) \end{bmatrix} \\ &= \begin{bmatrix} cD_1 f_i(a) & cD_2 f_i(a) & \dots & cD_n f_i(a) \end{bmatrix} = cDf_i(a) \end{aligned}$$



- (c) For  $f, g: X \subset \mathbb{R}^n \rightarrow \mathbb{R}$ , if  $f, g$  are differentiable at  $a \in X$ , then:  
 $fg$  is differentiable at  $a$  where  $D(fg)(a) = Df(a)g(a) + f(a)Dg(a)$

**Proof**

Since  $f, g$  are differentiable at  $a \in X$ :

$$\begin{aligned} D(fg)(a) &= \begin{bmatrix} D_1(fg)(a) & D_2(fg)(a) & \dots & D_n(fg)(a) \end{bmatrix} \\ &= \begin{bmatrix} D_1f(a)g(a) + f(a)D_1g(a) & D_2f(a)g(a) + f(a)D_2g(a) & \dots & D_nf(a)g(a) + f(a)D_ng(a) \end{bmatrix} \\ &= \begin{bmatrix} D_1f(a)g(a) & D_2f(a)g(a) & \dots & D_nf(a)g(a) \end{bmatrix} + \begin{bmatrix} f(a)D_1g(a) & f(a)D_2g(a) & \dots & f(a)D_ng(a) \end{bmatrix} \\ &= Df(a)g(a) + f(a)Dg(a) \end{aligned}$$

- (d) For  $f, g: X \subset \mathbb{R}^n \rightarrow \mathbb{R}$ , if  $f, g$  are differentiable at  $a \in X$  where  $g(a) \neq 0$ , then:  
 $\frac{f}{g}$  is differentiable at  $a$  where  $D(\frac{f}{g})(a) = \frac{Df(a)g(a) - f(a)Dg(a)}{[g(a)]^2}$

**Proof**

Since  $f, g$  are differentiable at  $a \in X$ :

$$\begin{aligned} D(\frac{f}{g})(a) &= \begin{bmatrix} D_1(\frac{f}{g})(a) & D_2(\frac{f}{g})(a) & \dots & D_n(\frac{f}{g})(a) \end{bmatrix} \\ &= \begin{bmatrix} \frac{D_1f(a)g(a) - f(a)D_1g(a)}{[g(a)]^2} & \frac{D_2f(a)g(a) - f(a)D_2g(a)}{[g(a)]^2} & \dots & \frac{D_nf(a)g(a) - f(a)D_ng(a)}{[g(a)]^2} \end{bmatrix} \\ &= \begin{bmatrix} \frac{D_1f(a)g(a)}{[g(a)]^2} & \frac{D_2f(a)g(a)}{[g(a)]^2} & \dots & \frac{D_nf(a)g(a)}{[g(a)]^2} \end{bmatrix} - \begin{bmatrix} \frac{f(a)D_1g(a)}{[g(a)]^2} & \frac{f(a)D_2g(a)}{[g(a)]^2} & \dots & \frac{f(a)D_ng(a)}{[g(a)]^2} \end{bmatrix} \\ &= Df(a) \frac{g(a)}{[g(a)]^2} - \frac{f(a)}{[g(a)]^2} Dg(a) = \frac{Df(a)g(a) - f(a)Dg(a)}{[g(a)]^2} \end{aligned}$$

### Definition 2.3.7: Partial Derivatives of Higher Orders

The **second order partial derivative** of  $f$  in respect to  $x_i$ :

$$\frac{\partial^2 f}{\partial x_i^2} = \frac{\partial}{\partial x_i} \left( \frac{\partial f}{\partial x_i} \right) = f_{x_i x_i}(x) = \lim_{h \rightarrow 0} \frac{f_{x_i}(x_1, \dots, x_i + h, \dots, x_n) - f_{x_i}(x_1, \dots, x_n)}{h}$$

The **mixed partial derivative** of  $f$  in respect to first  $x_i$ , then  $x_j$ :

$$\frac{\partial^2 f}{\partial x_j \partial x_i} = \frac{\partial}{\partial x_j} \left( \frac{\partial f}{\partial x_i} \right) = f_{x_i x_j}(x) = \lim_{h \rightarrow 0} \frac{f_{x_i}(x_1, \dots, x_j + h, \dots, x_n) - f_{x_i}(x_1, \dots, x_n)}{h}$$

In general, for  $f: X \subset \mathbb{R}^n \rightarrow \mathbb{R}$ , the  $k$ -th order partial derivative of  $f$  in respect to  $x_{i_1}, \dots, x_{i_k}$  in such order for  $k = \{1, \dots, n\}$ :

$$\begin{aligned} \frac{\partial^k f}{\partial x_{i_k} \dots \partial x_{i_1}} &= \frac{\partial}{\partial x_{i_k}} \dots \frac{\partial f}{\partial x_{i_1}} = f_{x_{i_1} \dots x_{i_k}}(x) \\ &= \lim_{h \rightarrow 0} \frac{f_{x_{i_1} \dots x_{i_{k-1}}}(x_1, \dots, x_{i_k} + h, \dots, x_n) - f_{x_{i_1} \dots x_{i_{k-1}}}(x_1, \dots, x_n)}{h} \end{aligned}$$

### Definition 2.3.8: Smoothness

For  $k = \{1, \dots, n\}$ ,  $f: X \subset \mathbb{R}^n \rightarrow \mathbb{R}$  is  $C^k$  if all partial derivatives of order 1 to  $k$  exist and are continuous on  $X$ .

If  $f$  has continuous partial derivatives of all order, then  $f$  is **smooth** (i.e  $C^\infty$ ).

For  $f: X \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$  where  $f = (f_1, \dots, f_m)$ , then  $f$  is  $C^k$  if each  $f_i$  is  $C^k$  for  $i = \{1, \dots, m\}$ .

**Theorem 2.3.9: Clairaut's Theorem**

If  $f: X \subset \mathbb{R}^n \rightarrow \mathbb{R}$  is  $C^k$ , then:

$$\frac{\partial^k f}{\partial x_{i_1} \dots \partial x_{i_k}} = \frac{\partial^k f}{\partial x_{j_1} \dots \partial x_{j_k}}$$

**Proof**

If the claim holds true for  $C^2$ , then replace  $f$  with  $f_{x_{i_p}}$  for any  $p = \{1, \dots, k\}$  and since  $f$  is  $C^k$ , then  $f_{x_{i_p}}$  is  $C^{k-1}$  and apply the theorem again. Repeating the process  $k$  times, the result holds true by induction. Now the proof for  $C^2$ :

Since  $f$  is  $C^2$ , then  $f_x, f_y, f_{xy}, f_{yx}$  exist and are continuous.

Let  $d(x, y) = f(x + h_1, y + h_2) - f(x + h_1, y) - f(x, y + h_2) + f(x, y)$ .

Since  $f_x$  exist, then by the Mean Value Theorem, there is a  $t_1 \in (0, h_1)$  where:

$$d(x, y) = h_1 * (f_x(x + t_1, y + h_2) - f_x(x + t_1, y))$$

Since  $f_y$  exist, then by the Mean Value Theorem, there is a  $t_2 \in (0, h_2)$  where:

$$d(x, y) = h_1 * h_2 * f_{xy}(x + t_1, y + t_2)$$

Since  $f_{xy}$  is continuous, then since  $(t_1, t_2) \rightarrow (0, 0)$  as  $(h_1, h_2) \rightarrow (0, 0)$ :

$$\begin{aligned} f_{xy}(x, y) &= \lim_{(h_1, h_2) \rightarrow (0, 0)} f_{xy}(x + h_1, x + h_2) \\ &= \lim_{(h_1, h_2) \rightarrow (0, 0)} f_{xy}(x + t_1, x + t_2) = \lim_{(h_1, h_2) \rightarrow (0, 0)} \frac{d(x, y)}{h_1 h_2} \end{aligned}$$

Rearrange  $d(x, y) = f(x + h_1, y + h_2) - f(x + h_1, y) - f(x, y + h_2) + f(x, y)$ .

Since  $f_y$  exist, by the Mean Value Theorem, there is a  $s_2 \in (0, h_2)$  where:

$$d(x, y) = h_2 * (f_y(x + h_1, y + s_2) - f_y(x, y + s_2))$$

Since  $f_x$  exist, by the Mean Value Theorem, there is a  $s_1 \in (0, h_1)$  where:

$$d(x, y) = h_2 * h_1 * f_{yx}(x + s_1, y + s_2)$$

Since  $f_{yx}$  is continuous, then since  $(s_1, s_2) \rightarrow (0, 0)$  as  $(h_1, h_2) \rightarrow (0, 0)$ :

$$\begin{aligned} f_{yx}(x, y) &= \lim_{(h_1, h_2) \rightarrow (0, 0)} f_{yx}(x + h_1, x + h_2) \\ &= \lim_{(h_1, h_2) \rightarrow (0, 0)} f_{xy}(x + s_1, x + s_2) = \lim_{(h_1, h_2) \rightarrow (0, 0)} \frac{d(x, y)}{h_1 h_2} \end{aligned}$$

Thus,  $f_{xy}(x, y) = f_{yx}(x, y)$ .

**Theorem 2.3.10: Chain Rule**

Let  $f: X \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$  be differentiable at  $x_0 \in X$  and  $g: f(X) \subset Y \subset \mathbb{R}^m \rightarrow \mathbb{R}^k$  be differentiable at  $f(x_0)$ .

Then  $g \circ f = g(f): X \subset \mathbb{R}^n \rightarrow \mathbb{R}^k$  is differentiable at  $x_0$  such that:

$$D[g(f(x_0))] = Dg(f(x_0)) Df(x_0)$$

**Proof**

Since  $f$  is differentiable at  $x_0$  and  $g$  is differentiable at  $f(x_0)$ , then there is a  $A = Df(x_0)$  and  $B = Dg(f(x_0))$  such that:

$$f(x_0 + h) - f(x_0) = Ah + r_A(h) \quad \text{where } \lim_{h \rightarrow 0} \frac{|r_A(h)|}{|h|} = 0$$

$$g(f(x_0) + k) - g(f(x_0)) = Bk + r_B(k) \quad \text{where } \lim_{k \rightarrow 0} \frac{|r_B(k)|}{|k|} = 0$$

Let  $k = f(x_0 + h) - f(x_0)$ . Thus:

$$\begin{aligned} g(f(x_0 + h)) - g(f(x_0)) &= BAh \\ &= g(f(x_0) + k) - g(f(x_0)) - BAh = Bk + r_B(k) - BAh = B(k - Ah) + r_B(k) \\ &= B(f(x_0 + h) - f(x_0) - Ah) + r_B(k) = Br_A(h) + r_B(k) \end{aligned}$$

Since  $f$  is differentiable at  $x_0$ , then  $f$  is continuous at  $x_0$  and thus,  $\lim_{h \rightarrow 0} k = 0$ .

Since  $\lim_{h \rightarrow 0} \frac{|r_A(h)|}{|h|} = 0$  and  $\lim_{k \rightarrow 0} \frac{|r_B(k)|}{|k|} = 0$ , then:

$$\lim_{h \rightarrow 0} \frac{|g(f(x_0 + h)) - g(f(x_0)) - BAh|}{|h|} \leq \lim_{h \rightarrow 0} (|B| \frac{|r_A(h)|}{|h|} + \frac{|r_B(k)|}{|h|}) = 0 + 0 = 0$$

Thus,  $D[g(f(x_0))] = BA = Dg(f(x_0)) Df(x_0)$ .

**Example**

Let  $f(u,v) = (u+v, uv, v^2)$  and  $g(x,y,z) = (x+y, \frac{z}{y}, x-y)$ . Find  $D(g \circ f)(u,v)$ .

By substitution:

$$\begin{aligned} D(g \circ f)(u,v) &= D(u+v+uv, \frac{v}{u}, u+v-uv) \\ &= \begin{bmatrix} \frac{\partial}{\partial u}u+v+uv & \frac{\partial}{\partial v}u+v+uv \\ \frac{\partial}{\partial u}\frac{v}{u} & \frac{\partial}{\partial v}\frac{v}{u} \\ \frac{\partial}{\partial u}u+v-uv & \frac{\partial}{\partial v}u+v-uv \end{bmatrix} = \begin{bmatrix} 1+v & 1+u \\ \frac{-v}{u^2} & \frac{1}{u} \\ 1-v & 1-u \end{bmatrix} \end{aligned}$$

Using the Chain Rule:

$$\begin{aligned} Dg(f(u,v)) Df(u,v) &= \begin{bmatrix} \frac{\partial}{\partial x}x+y & \frac{\partial}{\partial y}x+y & \frac{\partial}{\partial z}x+y \\ \frac{\partial}{\partial x}\frac{z}{y} & \frac{\partial}{\partial y}\frac{z}{y} & \frac{\partial}{\partial z}\frac{z}{y} \\ \frac{\partial}{\partial x}x-y & \frac{\partial}{\partial y}x-y & \frac{\partial}{\partial z}x-y \end{bmatrix} \begin{bmatrix} \frac{\partial}{\partial u}u+v & \frac{\partial}{\partial v}u+v \\ \frac{\partial}{\partial u}uv & \frac{\partial}{\partial v}uv \\ \frac{\partial}{\partial u}v^2 & \frac{\partial}{\partial v}v^2 \end{bmatrix} \\ &= \begin{bmatrix} 1 & 1 & 0 \\ 0 & \frac{-z}{y^2} & \frac{1}{y} \\ 1 & -1 & 0 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ v & u \\ 0 & 2v \end{bmatrix} = \begin{bmatrix} 1 & 1 & 0 \\ 0 & \frac{-1}{u^2} & \frac{1}{uv} \\ 1 & -1 & 0 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ v & u \\ 0 & 2v \end{bmatrix} = \begin{bmatrix} 1+v & 1+u \\ \frac{-v}{u^2} & \frac{1}{u} \\ 1-v & 1-u \end{bmatrix} \end{aligned}$$

**Theorem 2.3.11: Relationship between rectangular and polar partials**

For  $(x,y) = (r \cos(\theta), r \sin(\theta))$ :

$$\begin{aligned} \frac{\partial}{\partial r} &= \cos(\theta) \frac{\partial}{\partial x} + \sin(\theta) \frac{\partial}{\partial y} \\ \frac{\partial}{\partial \theta} &= -r \sin(\theta) \frac{\partial}{\partial x} + r \cos(\theta) \frac{\partial}{\partial y} \end{aligned}$$

Thus:

$$\begin{aligned} \frac{\partial}{\partial x} &= \cos(\theta) \frac{\partial}{\partial r} - \frac{\sin(\theta)}{r} \frac{\partial}{\partial \theta} \\ \frac{\partial}{\partial y} &= \sin(\theta) \frac{\partial}{\partial r} + \frac{\cos(\theta)}{r} \frac{\partial}{\partial \theta} \end{aligned}$$

**Proof**

Let  $z = g(r, \theta)$ . Then let  $z = f(x,y)$  such that  $(r \cos(\theta), r \sin(\theta)) = (x,y)$ .

By **theorem 2.3.10**:

$$\begin{aligned} D[g(r, \theta)] &= D(f(x, y)) D(x(r, \theta), y(r, \theta)) \\ \begin{bmatrix} \frac{\partial z}{\partial r} & \frac{\partial z}{\partial \theta} \end{bmatrix} &= \begin{bmatrix} \frac{\partial z}{\partial x} & \frac{\partial z}{\partial y} \end{bmatrix} \begin{bmatrix} \frac{\partial x}{\partial r} & \frac{\partial x}{\partial \theta} \\ \frac{\partial y}{\partial r} & \frac{\partial y}{\partial \theta} \end{bmatrix} = \begin{bmatrix} \frac{\partial z}{\partial x} & \frac{\partial z}{\partial y} \end{bmatrix} \begin{bmatrix} \cos(\theta) & -r \sin(\theta) \\ \sin(\theta) & r \cos(\theta) \end{bmatrix} \\ &= \begin{bmatrix} \cos(\theta) \frac{\partial z}{\partial x} + \sin(\theta) \frac{\partial z}{\partial y} & -r \sin(\theta) \frac{\partial z}{\partial x} + r \cos(\theta) \frac{\partial z}{\partial y} \end{bmatrix} \end{aligned}$$

Thus:

$$\begin{aligned} \frac{\partial}{\partial r} &= \cos(\theta) \frac{\partial}{\partial x} + \sin(\theta) \frac{\partial}{\partial y} \\ \frac{\partial}{\partial \theta} &= -r \sin(\theta) \frac{\partial}{\partial x} + r \cos(\theta) \frac{\partial}{\partial y} \end{aligned}$$

Then:

$$\begin{aligned} -r \cos(\theta) \frac{\partial}{\partial r} + \sin(\theta) \frac{\partial}{\partial \theta} &= -r \cos^2(\theta) \frac{\partial}{\partial x} - r \sin^2(\theta) \frac{\partial}{\partial x} = -r \frac{\partial}{\partial x} \\ r \sin(\theta) \frac{\partial}{\partial r} + \cos(\theta) \frac{\partial}{\partial \theta} &= r \sin^2(\theta) \frac{\partial}{\partial y} + r \cos^2(\theta) \frac{\partial}{\partial y} = r \frac{\partial}{\partial y} \end{aligned}$$

## 2.4 Directional Derivative

### Definition 2.4.1: Directional Derivative

Let  $f: X \subset \mathbb{R}^n \rightarrow \mathbb{R}$  be differentiable at  $a \in X$ . Then the **directional derivative** of  $f$  at  $a$  in the direction of vector  $v \in \mathbb{R}^n$ :

$$D_v f(a) = \lim_{h \rightarrow 0} \frac{f(a+hv) - f(a)}{\|hv\|}$$

### Theorem 2.4.2: Relationship between Directional Derivative and Gradient

Let  $f: X \subset \mathbb{R}^n \rightarrow \mathbb{R}$  be differentiable at  $a \in X$ . Then the directional derivative of  $f$  at  $a$  in the direction of vector  $v \in \mathbb{R}^n$ :

$$D_v f(a) = \nabla f(a) \cdot \frac{v}{\|v\|}$$

If  $v$  is a unit vector, then  $D_v f(a) = \nabla f(a) \cdot v$ .

### Proof

Let  $y(t) = a + tv$  for  $t \in (-\infty, \infty)$ . Then by **theorem 2.3.10**:

$$\begin{aligned} D_v f(a) &= \lim_{h \rightarrow 0} \frac{f(a+hv) - f(a)}{\|hv\|} = \lim_{h \rightarrow 0} \frac{f(y(h)) - f(y(0))}{|h|} \frac{1}{\|v\|} = Df(y(0)) Dy(0) \frac{1}{\|v\|} \\ &= Df(a) \frac{v}{\|v\|} = [Df(a)]^T \cdot \frac{v}{\|v\|} = \nabla f(a) \cdot \frac{v}{\|v\|} \end{aligned}$$

### Example

Let  $f(x,y) = 3xe^y$ . Find the derivative at  $(1,0)$  the direction of  $(1,1)$ .

$$\begin{aligned} \nabla f(1,0) &= (3e^y, 3xe^y)_{(x,y)=(1,0)} = (3,3) \\ D_{(1,1)} f(1,0) &= (3,3) \cdot \frac{1}{\sqrt{1^2+1^2}}(1,1) = \frac{6}{\sqrt{2}} = 3\sqrt{2} \end{aligned}$$

### Theorem 2.4.3: Direction of Steepest Ascent

The directional derivative  $D_v f(a) = \nabla f(a) \cdot \frac{v}{\|v\|}$  is:

Maximized when  $v$  is in the same direction as  $\nabla f(a)$  with value  $\|\nabla f(a)\|$

Minimized when  $v$  is in the opposite direction of  $\nabla f(a)$  with value  $-\|\nabla f(a)\|$

### Proof

By **theorem 1.2.3**,  $D_v f(a) = \nabla f(a) \cdot \frac{v}{\|v\|} = \|\nabla f(a)\| \left\| \frac{v}{\|v\|} \right\| \cos(\theta) = \|\nabla f(a)\| \cos(\theta)$ , where  $\theta \in [0, \pi]$  is the angle between  $\nabla f(a)$  and  $\frac{v}{\|v\|}$ .

Since  $D_v f(a)$  is maximized at  $\|\nabla f(a)\|$  when  $\theta = 0$ , then  $\nabla f(a)$  and  $v$  points in the same direction. Also,  $D_v f(a)$  is minimized at  $-\|\nabla f(a)\|$  when  $\theta = \pi$ , then  $\nabla f(a)$  and  $v$  points in opposite directions.

### Definition 2.4.4: Contour Curves / Level Curves

For  $f: X \subset \mathbb{R}^2 \rightarrow \mathbb{R}$ :

The **contour curve** at height  $c$  of  $f$  is the curve in  $\mathbb{R}^2$  where  $f(x,y) = c$ .

The **level curve** at height  $c$  of  $f$  is the same curve  $f(x,y) = c$ , but in  $\mathbb{R}^3$ .

**Theorem 2.4.5: Gradient is orthogonal to the surface**

If  $f: X \subset \mathbb{R}^n \rightarrow \mathbb{R}$  is  $C^1$ , then for any  $x_0$  where  $f(x_0) = c$  for constant  $c \in \mathbb{R}$ , then  $\nabla f(x_0)$  is orthogonal to the surface  $f(x) = c$  at  $x_0$ .

**Proof**

For surface  $f(x) = c$ , let curve  $C(t) = (x_1(t), \dots, x_n(t))$  where  $C(0) = x_0$  be defined such that  $f(C(t)) = c$ . Thus,  $\frac{d}{dt} f(C(t)) = \frac{d}{dt} c = 0$ .  
 Let  $v$  be the tangent vector to  $C(t)$  at  $x_0$ . Then by **theorem 2.3.10**, for  $t = 0$ :  

$$\frac{d}{dt} f(C(0)) = Df(C(0)) C'(0) = \nabla f(C(0)) \cdot C'(0) = \nabla f(x_0) \cdot v$$
  
 Since  $\nabla f(x_0) \cdot v = 0$  where  $v$  is tangent to  $C(t)$  which lies on surface  $f(x) = c$  and thus, is tangent to  $f(x) = c$ , then  $\nabla f(x_0)$  is orthogonal to  $f(x) = c$  at  $x_0$ .

### 3 Extrema

#### 3.1 Taylor's Theorem

##### Theorem 3.1.1: Taylor's Theorem for one variable

Let  $f: X \subset \mathbb{R} \rightarrow \mathbb{R}$  be  $C^{k+1}$ .

Then for  $a \in X$ , the  $k$ -th order Taylor polynomial of  $f$  centered at  $a$ :

$$p_k(x) = f(a) + f'(a)(x-a) + \frac{f''(a)}{2!}(x-a)^2 + \dots + \frac{f^{(k)}(a)}{k!}(x-a)^k$$

Then for  $h > 0$ , there is a  $x \in (a, a+h)$  such that:

$$f(a+h) = p_k(a+h) + \frac{f^{(k+1)}(x)}{(k+1)!}h^{k+1}$$

##### Proof

Let  $M$  be defined such that  $f(a+h) = p_k(a+h) + M(a+h-a)^{k+1}$ .

Let  $g(t) = f(t) - p_k(t) - M(t-a)^{k+1}$  for  $t \in [a, a+h]$ . Then:

$$g^{(k+1)}(t) = f^{(k+1)}(t) - (k+1)!M$$

Since  $p_k^{(i)}(a) = f^{(i)}(a)$  for  $i = \{0, \dots, k\}$ , then:

$$g^{(i)}(a) = f^{(i)}(a) - p_k^{(i)}(a) - \frac{(k+1)!}{(k+1-i)!}M(a-a)^{k+1-i} = 0$$

Since  $g(a+h) = 0$  where  $f$  is differentiable so  $g$  is differentiable, then by the Mean Value Theorem, there is a  $t_1 \in (0, h)$ :

$$0 - 0 = g(a+h) - g(a) = h * g'(a+t_1) \Rightarrow g'(a+t_1) = 0$$

Since  $g'$  is differentiable, by the Mean Value Theorem, there is a  $t_2 \in (0, t_1)$ :

$$0 - 0 = g'(a+t_1) - g'(a) = t_1 * g''(a+t_2) \Rightarrow g''(a+t_2) = 0$$

Repeating the process  $k+1$  times, there is a  $t_{k+1} \in (0, t_k)$ :

$$0 - 0 = g^{(k)}(a+t_k) - g^{(k)}(a) = t_k * g^{(k+1)}(a+t_{k+1}) \Rightarrow g^{(k+1)}(a+t_{k+1}) = 0$$

Since  $t_{k+1} < t_k < \dots < t_1 < h$ , then:

$$0 = g^{(k+1)}(a+t_{k+1}) = f^{(k+1)}(a+t_{k+1}) - (k+1)!M$$

Thus,  $M = \frac{f^{(k+1)}(a+t_{k+1})}{(k+1)!}$  where  $a+t_{k+1} \in [a, a+h]$ .

##### Theorem 3.1.2: Taylor's Theorem for multiple variables

Let  $f: X \subset \mathbb{R}^n \rightarrow \mathbb{R}$  be  $C^{k+1}$ . Then for  $a \in X$ :

$$p_k(x) = f(a) + \sum_{i=1}^n f_{x_i}(a)(x_i - a_i) + \frac{1}{2} \sum_{i,j=1}^n f_{x_i x_j}(a)(x_i - a_i)(x_j - a_j) + \dots + \frac{1}{k!} \sum_{i_1, \dots, i_k=1}^n f_{x_{i_1} \dots x_{i_k}}(a)(x_{i_1} - a_{i_1}) \dots (x_{i_k} - a_{i_k})$$

Then for  $h > 0$ :

$$f(a+h) = p_k(a+h) + R(a) \quad \text{where } \lim_{x \rightarrow a} \frac{R(a)}{\|x-a\|} = 0$$

##### Proof

Let  $g(t) = f(a+t(x-a))$ . Since  $f$  is  $C^{k+1}$ , then each  $D^i f$  for  $i = \{1, \dots, k+1\}$  exist and by [theorem 2.3.10](#):

$$g'(t) = Df(a+t(x-a))(x-a) = \sum_{i=1}^n f_{x_i}(a+t(x-a))(x_i - a_i)$$

$$g''(t) = \sum_{i=1}^n \sum_{j=1}^n f_{x_i x_j}(a+t(x-a))(x_j - a_j)(x_i - a_i)$$

$$g'''(t) = \sum_{i=1}^n \sum_{j=1}^n \sum_{k=1}^n f_{x_i x_j x_k}(a+t(x-a))(x_k - a_k)(x_j - a_j)(x_i - a_i)$$

etc

Since  $f$  is  $C^{k+1}$ , then  $g$  is  $C^{k+1}$  and by [theorem 3.1.1](#), there is a  $t \in (0, 1)$  such that:

$$g(1) = g(0) + g'(0)1 + \frac{g''(0)}{2!}1^2 + \dots + \frac{g^{(k)}(0)}{k!}1^k + \frac{g^{(k+1)}(t)}{(k+1)!}1^{k+1}$$

$$f(x) = f(a) + \sum_{i=1}^n f_{x_i}(a)(x_i - a_i) + \frac{1}{2!} \sum_{i=1}^n \sum_{j=1}^n f_{x_i x_j}(a)(x_j - a_j)(x_i - a_i) + \dots + \frac{1}{k!} \sum_{i_1, \dots, i_k=1}^n f_{x_{i_1} \dots x_{i_k}}(a)(x_{i_1} - a_{i_1}) \dots (x_{i_k} - a_{i_k}) + R(a)$$

**Example**

Let  $f(x,y,z) = e^x \sin(yz)$ .

Approximate  $f(0.1, \frac{\pi}{2} + 0.01, 1.05)$  using a 2nd degree Taylor Series.

$$\begin{aligned}
 f(0, \frac{\pi}{2}, 1) &= 1 \\
 Df(0, \frac{\pi}{2}, 1) &= \begin{bmatrix} e^x \sin(yz) & ze^x \cos(yz) & ye^x \cos(yz) \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \end{bmatrix} \\
 D^2f(0, \frac{\pi}{2}, 1) &= \begin{bmatrix} e^x \sin(yz) & ze^x \cos(yz) & ye^x \cos(yz) \\ ze^x \cos(yz) & -z^2 e^x \sin(yz) & -yze^x \sin(yz) \\ ye^x \cos(yz) & -yze^x \sin(yz) & -y^2 e^x \sin(yz) \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & -1 & -\frac{\pi}{2} \\ 0 & -\frac{\pi}{2} & -\frac{\pi^2}{4} \end{bmatrix} \\
 f(0.1, \frac{\pi}{2} + 0.01, 1.05) &\approx 1 + (1 * 0.1 + 0 * 0.01 + 0 * 0.05) \\
 &\quad + \frac{1}{2} (1 * 0.1^2 + 0 * 0.1 * 0.01 + 0 * 0.1 * 0.05 \\
 &\quad + 0 * 0.01 * 0.1 - 1 * 0.01^2 - \frac{\pi}{2} * 0.01 * 0.05 \\
 &\quad + 0 * 0.05 * 0.1 - \frac{\pi}{2} * 0.05 * 0.01 - \frac{\pi^2}{4} * 0.05^2) \approx 1.1011 \\
 \text{Actual } f(0.1, \frac{\pi}{2} + 0.01, 1.05) &\approx 1.1008 \text{ so the estimate is pretty close.}
 \end{aligned}$$

**3.2 Extrema****Definition 3.2.1: Local Extrema**

For  $f: X \subset \mathbb{R}^n \rightarrow \mathbb{R}$ , let  $a \in X$ .

$f$  has a **local minimum** at  $a$  if there is a  $\delta > 0$  such that:

$$f(x) \geq f(a) \text{ for all } x \in X \text{ where } \|x - a\| < \delta$$

$f$  has a **local maximum** at  $a$  if there is a  $\delta > 0$  such that:

$$f(x) \leq f(a) \text{ for all } x \in X \text{ where } \|x - a\| < \delta$$

**Theorem 3.2.2: Local extrema have a derivative of 0**

For differentiable  $f: X \subset \mathbb{R}^n \rightarrow \mathbb{R}$ , let  $a \in X$ .

If  $f$  has a local extrema at  $a$ , then  $Df(a) = 0$ .

**Proof**

Let  $f$  have a local maximum at  $a$ . The proof for local minimum is analogous.  
 For any  $i \in \{1, \dots, n\}$ , let  $F(t) = f(a + te_i)$ . Since  $F: \mathbb{R} \rightarrow \mathbb{R}$  has a local maximum at  $F(0)$ , then  $0 = F'(0) = \frac{\partial f}{\partial x_i}(a)$ .  
 Since  $Df(a) = \begin{bmatrix} \frac{\partial f}{\partial x_1}(a) & \frac{\partial f}{\partial x_2}(a) & \dots & \frac{\partial f}{\partial x_n}(a) \end{bmatrix}$ , then  $Df(a) = 0$ .

**Definition 3.2.3: Critical Point and Saddle Point**

For  $f: X \subset \mathbb{R}^n \rightarrow \mathbb{R}$ , let  $a \in X$ .

If  $Df(a) = 0$  or undefined, then  $a$  is a **critical point**.

Even if  $Df(a) = 0$ ,  $f(a)$  might not be a local extrema since it might be a local maximum along some paths and a local minimum along other paths. For this type of ambiguity of extrema,  $a$  is defined as a **saddle point**.

**Theorem 3.2.4: Second Derivative Test**

Let  $f: X \subset \mathbb{R}^2 \rightarrow \mathbb{R}$  be  $C^2$ .

If  $a \in X$  is a critical point of  $f$ , then:

- (a) If  $f_{xx}(a)f_{yy}(a) - [f_{xy}(a)]^2 > 0$  and  $f_{xx}(a) > 0$ , then  $a$  is a local minimum
- (b) If  $f_{xx}(a)f_{yy}(a) - [f_{xy}(a)]^2 > 0$  and  $f_{xx}(a) < 0$ , then  $a$  is a local maximum
- (c) If  $f_{xx}(a)f_{yy}(a) - [f_{xy}(a)]^2 < 0$ , then  $a$  is saddle point

**Example**

Let  $f(x,y) = 3x^2y + y^3 - 3x^2 - 3y^2 + 1$ . Find and label all critical points.

$$Df(x,y) = \begin{bmatrix} 6xy - 6x & 3x^2 + 3y^2 - 6y \end{bmatrix} = \begin{bmatrix} 0 & 0 \end{bmatrix}.$$

Thus,  $4x(y-1) = 0$  so  $x = 0$  or  $y = 1$ .

Thus,  $3x^2 + 3y^2 - 6y = 0$  for  $(x,y) = (0,0), (0,2), (1,1),$  and  $(-1,1)$ .

$$D^2f(x,y) = \begin{bmatrix} 6y - 6 & 6x \\ 6x & 6y - 6 \end{bmatrix}$$

At  $(0,0)$ ,  $\det(D^2f(0,0)) = 36$  and  $f_{xx}(0,0) = -6$  so local maximum.

At  $(0,2)$ ,  $\det(D^2f(0,0)) = 36$  and  $f_{xx}(0,0) = 6$  so local minimum.

At  $(1,1)$ ,  $\det(D^2f(0,0)) = -36$  so saddle point.

At  $(-1,1)$ ,  $\det(D^2f(0,0)) = -36$  so saddle point.

**Definition 3.2.5: Compactness**

Let set  $X \subset \mathbb{R}^n$ .

A point  $x \in X$  is a **limit point** if for any  $\delta > 0$ :

The set of all  $p$  where  $\|p - x\| < \delta$  contain a  $p \in X$  where  $p \neq x$ .

$X$  is closed if all limit point of  $X$  are in  $X$ .

$X$  is **bounded** if there is a  $M \in \mathbb{R}$  such that for all  $x \in X$ :

$$\|x\| < M$$

Then,  $X$  is **compact** if  $X$  is closed and bounded.

**Theorem 3.2.6: Extreme Value Theorem**

If  $X \subset \mathbb{R}^n$  is compact and  $f: X \rightarrow \mathbb{R}$  is continuous, then there are

$x_{\min}, x_{\max} \in X$  such that for all  $x \in X$ :

$$f(x_{\min}) \leq f(x) \leq f(x_{\max})$$

**Example**

Let  $f(x,y) = x^2 - xy + y^2 + 1$  where  $(x,y) \in [-2,2] \times [-2,2]$ .

Find the smallest and largest  $f(x,y)$  over  $[-2,2] \times [-2,2]$ .

For local extrema:

$$Df(x,y) = \begin{bmatrix} 2x - y & -x + 2y \end{bmatrix} = \begin{bmatrix} 0 & 0 \end{bmatrix} \text{ so } (x,y) = (0,0) \text{ with } f(0,0) = 1.$$

For boundary points, take the lines:

(a)  $x = -2$  for  $y \in [-2,2]$

$$f(-2,y) = 5 + 2y + y^2 = (y + 1)^2 + 4 \text{ so } (x,y) = (-2,-1) \text{ with } f(-2,-1) = 4$$

(b)  $x = 2$  for  $y \in [-2,2]$

$$f(2,y) = 5 - 2y + y^2 = (y - 1)^2 + 4 \text{ so } (x,y) = (2,1) \text{ with } f(2,1) = 4$$

(c)  $y = -2$  for  $x \in [-2,2]$

$$f(x,-2) = x^2 + 2x + 5 = (x + 1)^2 + 4 \text{ so } (x,y) = (-1,-2) \text{ with } f(-1,-2) = 4$$

(d)  $y = 2$  for  $x \in [-2,2]$

$$f(x,2) = x^2 - 2x + 5 = (x - 1)^2 + 4 \text{ so } (x,y) = (1,2) \text{ with } f(1,2) = 4$$

For endpoints,  $f(-2,-2) = 5$ ,  $f(-2,2) = 13$ ,  $f(2,-2) = 13$ ,  $f(2,2) = 5$ .

$\min(f(x,y)) = 1$  at  $(x,y) = (0,0)$        $\max(f(x,y)) = 13$  at  $(x,y) = (-2,2), (2,-2)$



### 3.3 Lagrange Multipliers for Constrained Extrema

#### Theorem 3.3.1: Lagrange Multiplier: Optimization for one constraint

Let  $f, g: X \subset \mathbb{R}^n \rightarrow \mathbb{R}$  be  $C^1$  and  $g(x) = c$  for constant  $c \in \mathbb{R}$ .

If  $f(x)$  has an extrema at  $x_0$  where  $g(x_0) = c$  and  $\nabla g(x_0) \neq 0$ , then there is a scalar  $\lambda \in \mathbb{R}$  such that:

$$\nabla f(x_0) = \lambda \nabla g(x_0)$$

#### Example

Let  $f(x,y,z) = 2xy + 2yz + xz$  where  $xyz = 4$ . Find the minimum of  $f(x,y,z)$ .

$$\nabla f(x, y, z) = \begin{bmatrix} 2y + z \\ 2x + 2z \\ 2y + x \end{bmatrix} = \lambda \nabla g(x, y, z) = \begin{bmatrix} \lambda yz \\ \lambda xz \\ \lambda xy \end{bmatrix}$$

Then,  $z - x = \lambda y(z - x)$  so  $x = z$  or  $y = \frac{1}{\lambda}$

If  $x = z$ , then  $4x = \lambda x^2$  so  $x = 0$  or  $x = \frac{4}{\lambda}$ .

If  $x = 0$ , then  $2y + 0 = \lambda 0y = 0$  so  $y = 0$ . Thus,  $(x,y,z) = (0,0,0)$ .

If  $x = \frac{4}{\lambda}$ , then  $2y + \frac{4}{\lambda} = \lambda \frac{4}{\lambda} y = 4y$  so  $y = \frac{2}{\lambda}$ . Thus,  $(\frac{4}{\lambda}, \frac{2}{\lambda}, \frac{4}{\lambda})$ .

If  $y = \frac{1}{\lambda}$ , then  $\frac{2}{\lambda} + z = z$  which is impossible.

Since  $\nabla f(0,0,0) = 0$ , then  $(x,y,z) \neq (0,0,0)$  so the only  $(x,y,z) = (\frac{4}{\lambda}, \frac{2}{\lambda}, \frac{4}{\lambda})$ .

Thus,  $4 = xyz = \frac{4}{\lambda} \frac{2}{\lambda} \frac{4}{\lambda} = \frac{32}{\lambda^3}$  so  $\lambda = 2$ .

Thus,  $(x,y,z) = (2,1,2)$  and the minimum  $f(x,y,z) = 4 + 4 + 4 = 12$ .

#### Theorem 3.3.2: Lagrange Multiplier: Optimization for multiple constraints

Let  $f, g_1, \dots, g_k: X \subset \mathbb{R}^n \rightarrow \mathbb{R}$  be  $C^1$  where  $k < n$  and each  $g_i(x) = c_i$  for constants  $c_i \in \mathbb{R}$  for  $i = \{1, \dots, k\}$ .

If  $f(x)$  has an extrema at  $x_0$  where each  $g_i(x_0) = c_i$  and  $\nabla g_i(x_0) \neq 0$ , then there are scalars  $\lambda_1, \dots, \lambda_k \in \mathbb{R}$  such that:

$$\nabla f(x_0) = \lambda_1 \nabla g_1(x_0) + \lambda_2 \nabla g_2(x_0) + \dots + \lambda_k \nabla g_k(x_0)$$

## 4 Integration

### 4.1 Double Integral

#### Definition 4.1.1: Riemann Sum in $\mathbb{R}^2$

For region  $R = [a,b] \times [c,d]$ , a **partition**,  $P$ , consist:

$\{x_0, \dots, x_n\} \in [a,b]$  such that each  $x_i < x_{i+1}$  and  $\Delta x_i = x_i - x_{i-1}$

$\{y_0, \dots, y_n\} \in [c,d]$  such that each  $y_i < y_{i+1}$  and  $\Delta y_i = y_i - y_{i-1}$

Let  $\|P\| = \max(\Delta x_i, \Delta y_j)$  for  $i,j = \{1, \dots, n\}$ . Then for each  $[x_{i-1}, x_i] \times [y_{j-1}, y_j]$ , choose one point  $c_{ij}$ . Let **sample points**,  $C$ , consist of all  $c_{ij}$ .

Then the **Riemann sum** of  $f(x,y)$  for partition  $P$ , sample points  $C$ :

$$R(f,P,C) = \sum_{i,j=1}^n f(c_{ij}) \Delta x_i \Delta y_j$$

#### Definition 4.1.2: Double Integral

$f: R \subset \mathbb{R}^2 \rightarrow \mathbb{R}$  is **integrable** if:

$$\int \int_R f(x,y) \, dA = \int \int_R f(x,y) \, dx dy = \lim_{\|P\| \rightarrow 0} \sum_{i,j=1}^n f(c_{ij}) \Delta x_i \Delta y_j$$

exist. Then,  $\int \int_R f(x,y) \, dA$  is the **double integral** of  $f$  on  $R$ .

#### Theorem 4.1.3: Continuous functions in $\mathbb{R}^2$ are integrable

If  $f: R \subset \mathbb{R}^2 \rightarrow \mathbb{R}$  is continuous on region  $R$ , then  $\int \int_R f(x,y) \, dA$  exists.

#### Theorem 4.1.4: Piecewise continuous functions in $\mathbb{R}^2$ are integrable

Set  $X \subset \mathbb{R}^2$  has **measure zero** if for any  $\epsilon > 0$  and every  $x_i \in X$ , there are  $[a_i, b_i] \times [c_i, d_i]$  where  $x_i \in [a_i, b_i] \times [c_i, d_i]$  such that:

$$\sum_{x_i \in X} (b_i - a_i) * (d_i - c_i) < \epsilon$$

If  $f: R \subset \mathbb{R}^2 \rightarrow \mathbb{R}$  is bounded and the set  $S$  of discontinuities of  $f$  on  $R$  has measure zero, then  $\int \int_R f(x,y) \, dA$  exists.

#### Theorem 4.1.5: Fubini's Theorem in $\mathbb{R}^2$ : Double Integrals $\Rightarrow$ Iterated Integrals

Let  $f: R = [a,b] \times [c,d] \subset \mathbb{R}^2 \rightarrow \mathbb{R}$  be bounded and the set  $S$  of discontinuities of  $f$  on  $R$  have measure zero. If every line parallel to  $x=0$  or  $y=0$  contains finitely many points on  $S$ , then:

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

#### Theorem 4.1.6: Properties of the Integral in $\mathbb{R}^2$

(a) If  $f, g$  are integrable on  $R$ , then  $f+g$  is integrable on  $R$  where:

$$\int \int_R f+g \, dA = \int \int_R f \, dA + \int \int_R g \, dA$$

(b) If  $f$  is integrable on  $R$  and scalar  $c \in \mathbb{R}$ , then  $cf$  is integrable on  $R$  where:

$$\int \int_R cf \, dA = c \int \int_R f \, dA$$

(c) If  $f, g$  are integrable on  $R$  where  $f(x,y) \leq g(x,y)$  for all  $(x,y) \in R$ :

$$\int \int_R f \, dA \leq \int \int_R g \, dA$$

(d) If  $f$  is integrable on  $R$ , then  $|f|$  is integrable on  $R$  where:

$$|\int \int_R f \, dA| \leq \int \int_R |f| \, dA$$

**Definition 4.1.7: Elementary Region and the Extension of f in  $\mathbb{R}^2$** 

Set  $D \subset \mathbb{R}^2$  is an **elementary region** if:

(a) **Type 1**

For  $x \in [a,b]$ , there are continuous  $y_1(x), y_2(x)$  where  $y_1(x) \leq y \leq y_2(x)$ .  
 $D = \{ (x,y) \mid x \in [a,b], y \in [y_1(x), y_2(x)] \}$

(b) **Type 2**

For  $y \in [c,d]$ , there are continuous  $x_1(y), x_2(y)$  where  $x_1(y) \leq x \leq x_2(y)$ .  
 $D = \{ (x,y) \mid x \in [x_1(y), x_2(y)], y \in [c,d] \}$

(c) **Type 3**

Both Type 1 and Type 2

If  $f: D \rightarrow \mathbb{R}^2 \rightarrow \mathbb{R}$  is continuous and  $D$  is an elementary region, then the extension of  $f$ ,  $f^{ext}$ :

$$f^{ext}(x, y) = \begin{cases} f(x, y) & (x, y) \in D \\ 0 & (x, y) \notin D \end{cases}$$

Since  $f$  is continuous and thus, integrable and the discontinuities of  $f^{ext}$  exist only at the boundary of  $D$  which is a curve and thus, have measure zero, then  $f^{ext}$  is integrable.

**Theorem 4.1.8: Double Integral over a General Region**

If  $f: D \rightarrow \mathbb{R}^2 \rightarrow \mathbb{R}$  is continuous and  $D$  is an elementary region, then for any region  $R = [a,b] \times [c,d]$  such that  $D \subset R$ :

$$\int \int_D f(x,y) \, dA = \int \int_R f^{ext}(x, y) \, dA$$

Then if:

(a)  $D$  is a Type 1 elementary region

$$\int \int_D f(x,y) \, dA = \int_a^b \int_{y_1(x)}^{y_2(x)} f(x,y) \, dy \, dx$$

(b)  $D$  is a Type 2 elementary region

$$\int \int_D f(x,y) \, dA = \int_c^d \int_{x_1(y)}^{x_2(y)} f(x,y) \, dx \, dy$$

**Proof**

Suppose  $D$  is a Type 1 elementary region.

Let region  $R = [a,b] \times [c,d]$  be such that  $c \leq y_1(x) \leq y \leq y_2(x) \leq d$ .

Since  $f^{ext}(x, y) = 0$  for any  $(x,y) \notin D$ , then by **theorem 4.1.5**:

$$\int \int_D f(x,y) \, dA = \int \int_R f^{ext}(x, y) \, dA = \int_a^b \int_c^d f^{ext}(x, y) \, dy \, dx = \int_a^b \int_{y_1(x)}^{y_2(x)} f(x,y) \, dy \, dx$$

Proof is analogous for a Type 2 elementary region.

**Example**

Let  $f(x,y) = x+y$ . Find  $\int \int_D f(x,y) \, dA$  for  $(x,y) \in \text{Triangle}((0,0), (3,0), (0,4))$ .

$$\begin{aligned} \int \int_D f(x,y) \, dA &= \int_0^3 \int_0^{\frac{4}{3}x} x+y \, dy \, dx = \int_0^3 [xy + \frac{1}{2}y^2]_0^{\frac{4}{3}x} \, dx = \int_0^3 \frac{20}{9}x^2 \, dx = 20 \\ &= \int_0^4 \int_{\frac{3}{4}y}^3 x+y \, dx \, dy = \int_0^4 [\frac{1}{2}x^2 + xy]_{\frac{3}{4}y}^3 \, dy = \int_0^4 \frac{9}{2} + 3y - \frac{33}{32}y^2 \, dy = 20 \end{aligned}$$

**Definition 4.1.9: Area**

The **area** of a region  $R \subset \mathbb{R}^2$ :

$$A = \int \int_R 1 \, dA$$

## 4.2 Triple Integral

### Definition 4.2.1: Riemann Sum in $\mathbb{R}^3$

For region  $R = [a,b] \times [c,d] \times [e,f]$ , a **partition**,  $P$ , consist:

$$\{x_0, \dots, x_n\} \in [a,b] \text{ such that each } x_i < x_{i+1} \text{ and } \Delta x_i = x_i - x_{i-1}$$

$$\{y_0, \dots, y_n\} \in [c,d] \text{ such that each } y_i < y_{i+1} \text{ and } \Delta y_i = y_i - y_{i-1}$$

$$\{z_0, \dots, z_n\} \in [e,f] \text{ such that each } z_i < z_{i+1} \text{ and } \Delta z_i = z_i - z_{i-1}$$

Let  $\|P\| = \max(\Delta x_i, \Delta y_j, \Delta z_k)$  for  $i,j,k = \{1, \dots, n\}$ . Then for each  $[x_{i-1}, x_i] \times [y_{j-1}, y_j] \times [z_{k-1}, z_k]$ , choose one point  $c_{ijk}$ . Let **sample points**,  $C$ , consist of all  $c_{ijk}$ . Then the **Riemann sum** of  $f(x,y,z)$  for partition  $P$ , sample points  $C$ :

$$R(f,P,C) = \sum_{i,j,k=1}^n f(c_{ijk}) \Delta x_i \Delta y_j \Delta z_k$$

### Definition 4.2.2: Triple Integral

$f: R \subset \mathbb{R}^3 \rightarrow \mathbb{R}$  is **integrable** if:

$$\begin{aligned} \iiint_R f(x,y,z) dV &= \iiint_R f(x,y,z) dx dy dz \\ &= \lim_{\|P\| \rightarrow 0} \sum_{i,j,k=1}^n f(c_{ijk}) \Delta x_i \Delta y_j \Delta z_k \end{aligned}$$

exist. Then,  $\iiint_R f(x,y,z) dV$  is the **triple integral** of  $f$  on  $R$ .

### Theorem 4.2.3: Continuous functions in $\mathbb{R}^3$ are integrable

If  $f: R \subset \mathbb{R}^3 \rightarrow \mathbb{R}$  is continuous on region  $R$ , then  $\iiint_R f(x,y,z) dV$  exists.

### Theorem 4.2.4: Piecewise continuous functions in $\mathbb{R}^3$ are integrable

Set  $X \subset \mathbb{R}^3$  has measure zero if for any  $\epsilon > 0$  and every  $x_i \in X$ , there are  $[a_i, b_i] \times [c_i, d_i] \times [e_i, f_i]$  where  $x_i \in [a_i, b_i] \times [c_i, d_i] \times [e_i, f_i]$  such that:

$$\sum_{x_i \in X} (b_i - a_i) * (d_i - c_i) * (f_i - e_i) < \epsilon$$

If  $f: R \subset \mathbb{R}^3 \rightarrow \mathbb{R}$  is bounded and the set  $S$  of discontinuities of  $f$  on  $R$  has measure zero, then  $\iiint_R f(x,y,z) dV$  exists.

### Theorem 4.2.5: Fubini's Theorem in $\mathbb{R}^3$ : Triple Integrals $\Rightarrow$ Iterated Integrals

Let  $f: R = [a,b] \times [c,d] \times [e,f] \subset \mathbb{R}^3 \rightarrow \mathbb{R}$  be bounded and the set  $S$  of discontinuities of  $f$  on  $R$  have measure zero. If every line parallel to  $x=0$ ,  $y=0$ , or  $z=0$  contains finitely many points on  $S$ , then:

$$\begin{aligned} \iiint_R f(x,y,z) dV &= \int_e^f \int_c^d \int_a^b f(x,y,z) dx dy dz = \int_c^d \int_e^f \int_a^b f(x,y,z) dx dz dy \\ &= \int_e^f \int_a^b \int_c^d f(x,y,z) dy dx dz = \int_a^b \int_e^f \int_c^d f(x,y,z) dy dz dx \\ &= \int_c^d \int_a^b \int_e^f f(x,y,z) dz dx dy = \int_a^b \int_c^d \int_e^f f(x,y,z) dz dy dx \end{aligned}$$

### Theorem 4.2.6: Properties of the Integral in $\mathbb{R}^3$

(a) If  $f, g$  are integrable on  $R$ , then  $f+g$  is integrable on  $R$  where:

$$\iiint_R f+g dV = \iiint_R f dV + \iiint_R g dV$$

(b) If  $f$  is integrable on  $R$  and scalar  $c \in \mathbb{R}$ , then  $cf$  is integrable on  $R$  where:

$$\iiint_R cf dV = c \iiint_R f dV$$

(c) If  $f, g$  are integrable on  $R$  where  $f(x,y) \leq g(x,y)$  for all  $(x,y) \in R$ :

$$\iiint_R f dV \leq \iiint_R g dV$$

(d) If  $f$  is integrable on  $R$ , then  $|f|$  is integrable on  $R$  where:

$$|\iiint_R f dV| \leq \iiint_R |f| dV$$

**Definition 4.2.7: Elementary Region and the Extension of  $f$  in  $\mathbb{R}^3$** 

Set  $D \subset \mathbb{R}^3$  is an elementary region if:

**(a) Type 1**

For  $x \in [a, b]$ , there are continuous  $y_1(x), y_2(x)$  where  $y_1(x) \leq y \leq y_2(x)$  and continuous  $z_1(x, y), z_2(x, y)$  where  $z_1(x, y) \leq z \leq z_2(x, y)$ .

$$D = \{ (x, y, z) \mid x \in [a, b], y \in [y_1(x), y_2(x)], z \in [z_1(x, y), z_2(x, y)] \}$$

OR

For  $x \in [a, b]$ , there are continuous  $z_1(x), z_2(x)$  where  $z_1(x) \leq z \leq z_2(x)$  and continuous  $y_1(x, z), y_2(x, z)$  where  $y_1(x, z) \leq y \leq y_2(x, z)$ .

$$D = \{ (x, y, z) \mid x \in [a, b], y \in [y_1(x, z), y_2(x, z)], z \in [z_1(x), z_2(x)] \}$$

**(b) Type 2**

For  $y \in [c, d]$ , there are continuous  $x_1(y), x_2(y)$  where  $x_1(y) \leq x \leq x_2(y)$  and continuous  $z_1(x, y), z_2(x, y)$  where  $z_1(x, y) \leq z \leq z_2(x, y)$ .

$$D = \{ (x, y, z) \mid x \in [x_1(y), x_2(y)], y \in [c, d], z \in [z_1(x, y), z_2(x, y)] \}$$

OR

For  $y \in [c, d]$ , there are continuous  $z_1(y), z_2(y)$  where  $z_1(y) \leq z \leq z_2(y)$  and continuous  $x_1(y, z), x_2(y, z)$  where  $x_1(y, z) \leq x \leq x_2(y, z)$ .

$$D = \{ (x, y, z) \mid x \in [x_1(y, z), x_2(y, z)], y \in [c, d], z \in [z_1(y), z_2(y)] \}$$

**(c) Type 3**

For  $z \in [e, f]$ , there are continuous  $x_1(z), x_2(z)$  where  $x_1(z) \leq x \leq x_2(z)$  and continuous  $y_1(x, z), y_2(x, z)$  where  $y_1(x, z) \leq y \leq y_2(x, z)$ .

$$D = \{ (x, y, z) \mid x \in [x_1(z), x_2(z)], y \in [y_1(x, z), y_2(x, z)], z \in [e, f] \}$$

OR

For  $z \in [e, f]$ , there are continuous  $y_1(z), y_2(z)$  where  $y_1(z) \leq y \leq y_2(z)$  and continuous  $x_1(y, z), x_2(y, z)$  where  $x_1(y, z) \leq x \leq x_2(y, z)$ .

$$D = \{ (x, y, z) \mid x \in [x_1(y, z), x_2(y, z)], y \in [y_1(z), y_2(z)], z \in [e, f] \}$$

**(d) Type 4**

All Type 1, 2, and 3

If  $f: D \rightarrow \mathbb{R}^3 \rightarrow \mathbb{R}$  is continuous and  $D$  is an elementary region, then the extension of  $f$ ,  $f^{ext}$ :

$$f^{ext}(x, y, z) = \begin{cases} f(x, y, z) & (x, y, z) \in D \\ 0 & (x, y, z) \notin D \end{cases}$$

Since  $f$  is continuous and thus, integrable and the discontinuities of  $f^{ext}$  exist only at the boundary of  $D$  which is a 2d surface and thus, have measure zero, then  $f^{ext}$  is integrable.

**Theorem 4.2.8: Triple Integral over a General Region**

If  $f: D \rightarrow \mathbb{R}^3 \rightarrow \mathbb{R}$  is continuous and  $D$  is an elementary region, then for any region  $R = [a,b] \times [c,d] \times [e,f]$  such that  $D \subset R$ :

$$\iiint_D f(x,y,z) \, dV = \iiint_R f^{ext}(x,y,z) \, dV$$

Then if:

(a)  $D$  is a Type 1 elementary region

$$\begin{aligned} \iiint_D f(x,y,z) \, dV &= \int_a^b \int_{y_1(x)}^{y_2(x)} \int_{z_1(x,y)}^{z_2(x,y)} f(x,y,z) \, dz dy dx \\ &= \int_a^b \int_{z_1(x)}^{z_2(x)} \int_{y_1(x,z)}^{y_2(x,z)} f(x,y,z) \, dy dz dx \end{aligned}$$

(b)  $D$  is a Type 2 elementary region

$$\begin{aligned} \iiint_D f(x,y,z) \, dV &= \int_c^d \int_{x_1(y)}^{x_2(y)} \int_{z_1(x,y)}^{z_2(x,y)} f(x,y,z) \, dz dx dy \\ &= \int_c^d \int_{z_1(y)}^{z_2(y)} \int_{x_1(y,z)}^{x_2(y,z)} f(x,y,z) \, dx dz dy \end{aligned}$$

(c)  $D$  is a Type 3 elementary region

$$\begin{aligned} \iiint_D f(x,y,z) \, dV &= \int_e^f \int_{x_1(z)}^{x_2(z)} \int_{y_1(x,z)}^{y_2(x,z)} f(x,y,z) \, dy dx dz \\ &= \int_e^f \int_{y_1(z)}^{y_2(z)} \int_{x_1(y,z)}^{x_2(y,z)} f(x,y,z) \, dx dy dz \end{aligned}$$

**Proof**

Suppose  $D$  is a Type 1 elementary region.

Let region  $R = [a,b] \times [c,d] \times [e,f]$  be such that  $c \leq y_1(x) \leq y \leq y_2(x) \leq d$  and  $e \leq z_1(x,y) \leq z \leq z_2(x,y) \leq f$ .

Since  $f^{ext}(x,y,z) = 0$  for any  $(x,y,z) \notin D$ , then by **theorem 4.1.5**:

$$\begin{aligned} \iiint_D f(x,y,z) \, dV &= \iiint_R f^{ext}(x,y,z) \, dV = \int_a^b \int_c^d \int_e^f f^{ext}(x,y,z) \, dz dy dx \\ &= \int_a^b \int_c^d \int_{z_1(x,y)}^{z_2(x,y)} f^{ext}(x,y,z) \, dz dy dx \\ &= \int_a^b \int_{y_1(x)}^{y_2(x)} \int_{z_1(x,y)}^{z_2(x,y)} f(x,y,z) \, dz dy dx \end{aligned}$$

Proof is analogous for the other five cases.

**Example**

Let  $f(x,y,z) = x^2y + z$ . Find  $\iiint_D f(x,y,z) \, dV$  for  $(x,y,z) \in \text{TriangularPyramid}((0,0,0), (3,0,0), (0,4,0), (3,4,5))$ .

$$\begin{aligned} \iiint_D f(x,y,z) \, dV &= \int_0^3 \int_0^{\frac{4}{3}x} \int_0^{\frac{5}{4}y} x^2y + z \, dz dy dx \\ &= \int_0^3 \int_0^{\frac{4}{3}x} \frac{5}{4}x^2y^2 + \frac{25}{32}y^2 \, dy dx = \int_0^3 \frac{80}{81}x^5 + \frac{50}{81}x^3 \, dx = \frac{265}{2} \end{aligned}$$

**Definition 4.2.9: Volume**

The **volume** of a region  $R \subset \mathbb{R}^3$ :

$$V = \iiint_R 1 \, dV$$

### 4.3 Change of Variables

#### Theorem 4.3.1: Linear Transformations Scaling Factor

Let  $A \in M_{2 \times 2}(\mathbb{R})$  where  $\det(A) \neq 0$  such that  $T: \mathbb{R}^2 \rightarrow \mathbb{R}^2$ :

$$(x,y) = T(u,v) = A \begin{bmatrix} u \\ v \end{bmatrix}$$

Transformation  $T$  is 1-1 and onto where if  $D$  is a parallelogram in the  $uv$ -plane, then  $T(D)$  is a parallelogram in the  $xy$ -plane where:

$$\text{Vol}_2(T(D)) = |\det(A)| * \text{Vol}_2(D)$$

#### Proof

The proof that  $T$  is 1-1 and onto uses linear algebra, but it will be made simpler here. Note however if linear algebra is used in full effect, the given  $\det(A) \neq 0$  satisfies the condition for 1-1 and onto.

Let  $A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$ . Suppose  $(a,c) = k(b,d)$  for some scalar  $k \in \mathbb{R}$ .

$$A \begin{bmatrix} u \\ v \end{bmatrix} = \begin{bmatrix} au + bv \\ cu + dv \end{bmatrix} = \begin{bmatrix} a \\ c \end{bmatrix} u + \begin{bmatrix} b \\ d \end{bmatrix} v = k \begin{bmatrix} b \\ d \end{bmatrix} u + \begin{bmatrix} b \\ d \end{bmatrix} v = \begin{bmatrix} b \\ d \end{bmatrix} (ku + v)$$

Let  $(u,v) = (1,-k)$  and  $(u,v) = (2,-2k)$ . In both cases,  $A \begin{bmatrix} u \\ v \end{bmatrix} = 0$  so  $T$  is not 1-1 if

$(a,c) = k(b,d)$ . But, if there is a  $k$  such that  $(a,c) = k(b,d)$ , then  $\det(A) = ad - bc = (kb)d - b(kd) = 0$  which contradicts that  $\det(A) \neq 0$ . Also, note if there is a  $k$  such that  $(a,c) = k(b,d)$ , then  $A(u,v)$  consist of only multiples of  $(b,d)$  so  $T$  is not onto since vectors that are not multiples of  $(b,d)$  such as  $(b,d+1)$  cannot be reached. Now suppose there isn't a  $k$  such that  $(a,c) = k(b,d)$ . Then  $A(u,v)$  cannot have two different  $(u_1, v_1), (u_2, v_2)$  such that  $A(u_1, v_1) = A(u_2, v_2)$  since

$$A(u_1, v_1) - A(u_2, v_2) = (a,c)(u_1 - u_2) + (b,d)(v_1 - v_2) = 0$$

so  $(a,c) = -\frac{v_1 - v_2}{u_1 - u_2}(b,d)$  and thus, there is a  $k = -\frac{v_1 - v_2}{u_1 - u_2}$  such that  $(a,c) = k(b,d)$  which is a contradiction. Thus, if there isn't a  $k$  such that  $(a,c) = k(b,d)$ , then for two different  $(u_1, v_1)$  and  $(u_2, v_2)$ , then  $A(u_1, v_1) \neq A(u_2, v_2)$  so  $T$  is 1-1. Also, note since there isn't a  $k$  such that  $(a,c) = k(b,d)$ , then any  $(x,y)$  can be reached by solving the system of equations from  $A(u,v) = (x,y)$  for  $u,v$  and thus,  $T$  is onto.

If there isn't a  $k$  such that  $(a,b) = k(c,d)$ , then  $\det(A) \neq 0$  since if  $\det(A) = ad - bc = 0$ , then  $ad = bc$  so  $\frac{a}{b} = \frac{c}{d}$ . Note  $\frac{a}{b} = \frac{kc}{kd} = \frac{c}{d}$  if there is a  $k$  such that  $(a,b) = k(c,d)$  which is a contradiction. Thus, by the restriction  $\det(A) \neq 0$ , then  $T$  is 1-1 and onto.

For any two  $(u_1, v_1), (u_2, v_2)$  where  $(u_1, v_1) \neq k(u_1, v_1)$  for any  $k \in \mathbb{R}$ , a parallelogram with sides  $(u_1, v_1), (u_2, v_2)$  can be formed in  $uv$ -plane. Then  $T$  maps  $(u_1, v_1), (u_2, v_2)$  into two different  $A(u_1, v_1), A(u_2, v_2)$  since  $T$  is 1-1 so  $A(u_1, v_1), A(u_2, v_2)$  are sides of a parallelogram in the  $xy$ -plane. Thus:

$$\begin{aligned} \text{Vol}_2(T(D)) &= \|A(u_1, v_1) \times A(u_2, v_2)\| = \|[(a, c)u_1 + (b, d)v_1] \times [(a, c)u_2 + (b, d)v_2]\| \\ &= \|(a, c)u_1 \times (a, c)u_2 + (a, c)u_1 \times (b, d)v_2 + (b, d)v_1 \times (a, c)u_2 + (b, d)v_1 \times (b, d)v_2\| \\ &= \|u_1v_2 - v_1u_2\| \|(a, c) \times (b, d)\| = \|(a, c) \times (b, d)\| |u_1v_2 - v_1u_2| \\ &= |ad - bc| \|(u_1, v_1) \times (u_2, v_2)\| = |\det(A)| * \text{Vol}_2(D) \end{aligned}$$

**Theorem 4.3.2: Integral: Change of Variables**

The **Jacobian** of  $T: \mathbb{R}^2 \rightarrow \mathbb{R}^2$  where  $T(u,v) = (x,y)$ :

$$\frac{\partial(x,y)}{\partial(u,v)} = \det(DT(u,v)) = \det\left(\begin{bmatrix} \frac{\partial x}{\partial u} & \frac{\partial x}{\partial v} \\ \frac{\partial y}{\partial u} & \frac{\partial y}{\partial v} \end{bmatrix}\right)$$

Let elementary region  $D$  be in  $uv$ -plane. If transformation  $T$  is  $C^1$  (i.e. partial derivatives above are continuous), then for integrable  $f: T(D) \rightarrow \mathbb{R}$  where  $x = x(u,v)$  and  $y = y(u,v)$ :

$$\int \int_{T(D)} f(x,y) \, dA(x,y) = \int \int_D f(x(u,v), y(u,v)) \left| \frac{\partial(x,y)}{\partial(u,v)} \right| \, dA(u,v)$$

The Jacobian of  $T: \mathbb{R}^3 \rightarrow \mathbb{R}^3$  where  $T(u,v,w) = (x,y,z)$ :

$$\frac{\partial(x,y,z)}{\partial(u,v,w)} = \det(DT(u,v)) = \det\left(\begin{bmatrix} \frac{\partial x}{\partial u} & \frac{\partial x}{\partial v} & \frac{\partial x}{\partial w} \\ \frac{\partial y}{\partial u} & \frac{\partial y}{\partial v} & \frac{\partial y}{\partial w} \\ \frac{\partial z}{\partial u} & \frac{\partial z}{\partial v} & \frac{\partial z}{\partial w} \end{bmatrix}\right)$$

Let elementary region  $D$  be in  $uvw$ -space. If transformation  $T$  is  $C^1$ , then for integrable  $f: T(D) \rightarrow \mathbb{R}$  where  $x = x(u,v,w)$ ,  $y = y(u,v,w)$ , and  $z = z(u,v,w)$ :

$$\begin{aligned} \int \int \int_{T(D)} f(x,y,z) \, dV(x,y,z) \\ = \int \int \int_D f(x(u,v,w), y(u,v,w), z(u,v,w)) \left| \frac{\partial(x,y,z)}{\partial(u,v,w)} \right| \, dV(u,v,w) \end{aligned}$$

**Proof**

Take the case for  $T: \mathbb{R}^2 \rightarrow \mathbb{R}^2$ . The case for  $\mathbb{R}^3$  is analogous.

Since  $T$  is  $C^1$ , then  $DT(u,v)$  exist and since  $T(u,v) = A(u,v) = (x(u,v), y(u,v))$ , then:

$$DT(u,v) = A' = \begin{bmatrix} \frac{\partial x}{\partial u} & \frac{\partial x}{\partial v} \\ \frac{\partial y}{\partial u} & \frac{\partial y}{\partial v} \end{bmatrix} = \left| \frac{\partial(x,y)}{\partial(u,v)} \right|$$

Thus, by **theorem 4.3.1**:

$$\begin{aligned} \int \int_{T(D)} f(x,y) \, dA(x,y) &= \lim_{\|P\| \rightarrow 0} \sum_{i,j=1}^n f(x_{ij}, y_{ij}) \Delta x_i \Delta y_j \\ &= \lim_{\|P\| \rightarrow 0} \sum_{i,j=1}^n f(x_{ij}, y_{ij}) \text{Vol}_2([\Delta x_i, \Delta y_j]) \\ &= \lim_{\|P\| \rightarrow 0} \sum_{i,j=1}^n f(x_{ij}(u,v), y_{ij}(u,v)) |DT(u,v)| \text{Vol}_2([\Delta u_i, \Delta v_j]) \\ &= \lim_{\|P\| \rightarrow 0} \sum_{i,j=1}^n f(x_{ij}(u,v), y_{ij}(u,v)) |DT(u,v)| \Delta u_i \Delta v_j \\ &= \int \int_D f(x(u,v), y(u,v)) \left| \frac{\partial(x,y)}{\partial(u,v)} \right| \, dA(u,v) \end{aligned}$$

**Example**

Prove that the volume of a sphere with radius  $R$  is  $V = \frac{4}{3}\pi R^3$ .

A sphere is  $x^2 + y^2 + z^2 = R^2$ . Instead of rectangular, use spherical coordinates.

$$p^2 = r^2 + z^2 = x^2 + y^2 + z^2 = R^2 \quad \Rightarrow \quad p = R$$

$\theta \in [0, 2\pi]$  since a sphere in 2d is a circle and  $\phi \in [0, \pi]$  since for  $(x,y) = (0,0)$ , then  $z = [-R, R]$  where  $z = p \cos(\phi) = R \cos(\phi)$ .

Since  $x = p \sin(\phi) \cos(\theta)$ ,  $y = p \sin(\phi) \sin(\theta)$ , and  $z = p \cos(\phi)$ , then:

$$\left| \frac{\partial(x,y,z)}{\partial(p,\theta,\phi)} \right| = \left| \begin{vmatrix} \sin(\phi) \cos(\theta) & -p \sin(\phi) \sin(\theta) & p \cos(\phi) \cos(\theta) \\ \sin(\phi) \sin(\theta) & p \sin(\phi) \cos(\theta) & p \cos(\phi) \sin(\theta) \\ \cos(\phi) & 0 & -p \sin(\phi) \end{vmatrix} \right| = |-p^2 \sin(\phi)| = p^2 \sin(\phi)$$

$$\begin{aligned} \text{Volume} &= \int \int \int_{x^2+y^2+z^2 \leq R^2} 1 \, dV(x,y,z) = \int_0^R \int_0^\pi \int_0^{2\pi} 1 * p^2 \sin(\phi) \, d\theta d\phi dp \\ &= 2\pi \int_0^R \int_0^\pi p^2 \sin(\phi) \, d\phi dp = 2\pi \int_0^R 2p^2 \, dp = \frac{4}{3}\pi R^3 \end{aligned}$$



## 5 Line Integrals

### 5.1 Parametrized Curves

#### Definition 5.1.1: Path

A **path** in  $\mathbb{R}^n$  is a continuous  $C(t): [a, b] \rightarrow \mathbb{R}^n$ .

If  $C(t)$  is twice-differentiable, then the **velocity** of  $C(t)$  at  $t_0 \in [a, b]$ :

$$v(t_0) = x'(t_0)$$

Also, the acceleration of  $C(t)$  at  $t_0 \in [a, b]$ :

$$a(t_0) = x''(t_0)$$

#### Definition 5.1.2: Arclength

The length of a  $C^1$  path  $C(t): [a, b] \rightarrow \mathbb{R}^n$ :

$$L = \int_a^b \|C'(t)\| dt$$

#### Proof

Choose  $\{x_1, \dots, x_n\} \in [a, b]$  such that each  $x_i < x_{i+1}$ . Then the length of  $C(t)$ :

$$L = \lim_{n \rightarrow \infty} \sum_{i=1}^n \|C(x_i) - C(x_{i-1})\|$$

Let each  $C(x_i) = (C_1(x_i), \dots, C_n(x_i))$ . Thus:

$$\|C(x_i) - C(x_{i-1})\| = \sqrt{(C_1(x_i) - C_1(x_{i-1}))^2 + \dots + (C_n(x_i) - C_n(x_{i-1}))^2}$$

Since  $C(t)$  is  $C^1$ , by the Mean Value Theorem, there is a  $t_{i_k} \in [x_{i+1}, x_i]$  such that:

$$C_k(x_i) - C_k(x_{i-1}) = (x_i - x_{i-1})C'_k(t_{i_k})$$

Thus:

$$\|C(x_i) - C(x_{i-1})\| = \sqrt{(x_i - x_{i-1})^2 [C'_1(t_{i_1})]^2 + \dots + (x_i - x_{i-1})^2 [C'_n(t_{i_n})]^2}$$

$$L = \lim_{n \rightarrow \infty} \sum_{i=1}^n \sqrt{[C'_1(t_{i_1})]^2 + \dots + [C'_n(t_{i_n})]^2} (x_i - x_{i-1})$$

$$= \int_a^b \sqrt{[C'_1(t)]^2 + \dots + [C'_n(t)]^2} dt = \int_a^b \|C'(t)\| dt$$

#### Example

Find the length of the helix  $C(t) = (5\cos(t), 5\sin(t), 12t)$  for  $t \in [0, 2\pi]$ .

$$C'(t) = (-5\sin(t), 5\cos(t), 12) \text{ so } \|C'(t)\| = \sqrt{25\sin^2(t) + 25\cos^2(t) + 144} = 13.$$

$$L = \int_0^{2\pi} 13 dt = 26\pi$$

### 5.2 Vector Fields

#### Definition 5.2.1: Vector Field and Flow Lines

A **vector field** on  $\mathbb{R}^n$  is  $F: X \subset \mathbb{R}^n \rightarrow \mathbb{R}^n$

A **flow line** of vector field  $F$  is a differentiable path  $C(t): [a, b] \rightarrow \mathbb{R}^n$  such that:

$$C'(t) = F(C(t))$$

#### Definition 5.2.2: Del Operator

The **Del Operator** on  $\mathbb{R}^n$ :

$$\nabla = \sum_{i=1}^n \frac{\partial}{\partial x_i} e_i = \left( \frac{\partial}{\partial x_1}, \dots, \frac{\partial}{\partial x_n} \right)$$

**Definition 5.2.3: Divergence**

For differentiable vector field  $F: X \subset \mathbb{R}^n \rightarrow \mathbb{R}^n$ , let  $F = (F_1, \dots, F_n)$ .

Then the **divergence**,  $\text{div}: \mathbb{R}^n \rightarrow \mathbb{R}$ , of  $F$ :

$$\text{div}(F) = \nabla \cdot F = \frac{\partial F_1}{\partial x_1} + \dots + \frac{\partial F_n}{\partial x_n}$$

If  $\text{div}(F) = 0$  everywhere, then  $F$  is incompressible.

**Definition 5.2.4: Curl**

For differentiable vector field  $F: X \subset \mathbb{R}^3 \rightarrow \mathbb{R}^3$ , let  $F = (F_1, F_2, F_3)$ .

Then the **curl**,  $\text{curl}: \mathbb{R}^n \rightarrow \mathbb{R}^n$ , of  $F$ :

$$\text{curl}(F) = \nabla \times F = \begin{bmatrix} e_1 & e_2 & e_3 \\ \frac{\partial}{\partial x} & \frac{\partial}{\partial y} & \frac{\partial}{\partial z} \\ F_1 & F_2 & F_3 \end{bmatrix} = \left( \frac{\partial F_3}{\partial y} - \frac{\partial F_2}{\partial z}, \frac{\partial F_1}{\partial z} - \frac{\partial F_3}{\partial x}, \frac{\partial F_2}{\partial x} - \frac{\partial F_1}{\partial y} \right)$$

If  $\text{curl}(F) = 0$  everywhere, then  $F$  is irrotational.

**Theorem 5.2.5: Vector fields from Gradients are irrotational**

If  $f: X \subset \mathbb{R}^3 \rightarrow \mathbb{R}$  is  $C^2$ , then  $\text{curl}(\nabla f) = 0$

**Proof**

Since  $\nabla f = \left( \frac{\partial f}{\partial x}, \frac{\partial f}{\partial y}, \frac{\partial f}{\partial z} \right)$ , then by **theorem 2.3.9**:

$$\text{curl}(\nabla f) = \begin{bmatrix} e_1 & e_2 & e_3 \\ \frac{\partial}{\partial x} & \frac{\partial}{\partial y} & \frac{\partial}{\partial z} \\ \frac{\partial f}{\partial x} & \frac{\partial f}{\partial y} & \frac{\partial f}{\partial z} \end{bmatrix} = \left( \frac{\partial^2 f}{\partial y \partial z} - \frac{\partial^2 f}{\partial z \partial y}, \frac{\partial^2 f}{\partial z \partial x} - \frac{\partial^2 f}{\partial x \partial z}, \frac{\partial^2 f}{\partial x \partial y} - \frac{\partial^2 f}{\partial y \partial x} \right) = 0$$

**Theorem 5.2.6: The curl is incompressible**

If  $F: X \subset \mathbb{R}^3 \rightarrow \mathbb{R}^3$  is  $C^2$ , then  $\text{div}(\text{curl}(F)) = 0$

**Proof**

Since  $\text{curl}(F) = \left( \frac{\partial F_3}{\partial y} - \frac{\partial F_2}{\partial z}, \frac{\partial F_1}{\partial z} - \frac{\partial F_3}{\partial x}, \frac{\partial F_2}{\partial x} - \frac{\partial F_1}{\partial y} \right)$ , then by **theorem 2.3.9**:

$$\text{div}(\text{curl}(F)) = \frac{\partial}{\partial x} \left( \frac{\partial F_3}{\partial y} - \frac{\partial F_2}{\partial z} \right) + \frac{\partial}{\partial y} \left( \frac{\partial F_1}{\partial z} - \frac{\partial F_3}{\partial x} \right) + \frac{\partial}{\partial z} \left( \frac{\partial F_2}{\partial x} - \frac{\partial F_1}{\partial y} \right) = 0$$

**5.3 Line Integrals****Theorem 5.3.1: Scalar Line Integral**

Let  $C^1$  path  $C(t): [a, b] \rightarrow \mathbb{R}^n$  and  $f: C([a, b]) \subset X \subset \mathbb{R}^n \rightarrow \mathbb{R}$  be continuous.

Then the **scalar line integral** of  $f$  along  $C(t)$ :

$$\int_{C(t)} f \, ds = \int_a^b f(C(t)) \|C'(t)\| \, dt$$

**Proof**

Let partition  $P = \{t_0, \dots, t_n\} \subset [a, b]$  with sample points  $C = \{t_1^*, \dots, t_n^*\}$ , the Riemann sum of  $f$  along  $C(t)$  with partition  $P$  and sample points  $C$ :

$$\sum_{i=1}^n f(C(t_i^*)) \Delta s_i \quad \text{where } \Delta s_i = \int_{t_{i-1}}^{t_i} \|C'(t)\| \, dt$$

Since  $C(t)$  is  $C^1$  so  $\Delta s_i$  is differentiable, then by the Mean Value Theorem, there is a  $t_i^{**} \in [t_{i-1}, t_i]$ :

$$\Delta s_i = \int_{t_{i-1}}^{t_i} \|C'(t)\| \, dt = (t_i - t_{i-1}) \|C'(t_i^{**})\| = \|C'(t_i^{**})\| \Delta t_i$$

Thus:

$$\lim_{\|P\| \rightarrow 0} \sum_{i=1}^n f(C(t_i^*)) \Delta s_i = \lim_{\|P\| \rightarrow 0} \sum_{i=1}^n f(C(t_i^*)) \|C'(t_i^{**})\| \Delta t_i = \int_a^b f(C(t)) \|C'(t)\| \, dt$$

**Example**

Let  $f(x,y) = xy$ .

Find the scalar line integral of  $f$  along  $C(t) = (t, t^2)$  for  $t \in [0,1]$

$$\int_{C(t)} f \, ds = \int_0^1 t^3 \sqrt{1+4t^2} \, dt = \left[ \frac{1}{32} \left( \frac{2}{5} (1+4t^2)^{\frac{5}{2}} - \frac{2}{3} (1+4t^2)^{\frac{3}{2}} \right) \right]_0^1 = \frac{5^{5/2}+1}{120}$$

**Definition 5.3.2: Vector Line Integral**

Let  $C^1$  path  $C(t): [a,b] \rightarrow \mathbb{R}^n$  and  $F: C([a,b]) \subset X \subset \mathbb{R}^n \rightarrow \mathbb{R}^n$  be continuous. Then the **vector line integral** of  $F$  along  $C(t)$ :

$$\int_{C(t)} F \cdot ds = \int_a^b F(C(t)) \cdot C'(t) \, dt$$

**Physical Understanding #1: Vector Line Integrals = Total Work**

In physics, Work = Force \* Change in distance in a direction  $d$

$$W = \|F\| \cos(\theta) \|\Delta x\| = \|F\| \|\Delta x\| \cos(\theta) = F \cdot \Delta x$$

Since  $x'(t_0) = \lim_{\Delta t \rightarrow 0} \frac{x(\Delta t + t_0) - x(t_0)}{\Delta t} = \lim_{\Delta t \rightarrow 0} \frac{\Delta x}{\Delta t}$ , then the work at  $t_0$ :

$$W \approx F(x(t_0)) \cdot x'(t_0) \Delta t$$

So the total work for all points  $t_0$  in a path:

$$\sum_{t_0 \in \text{path}} F(x(t_0)) \cdot x'(t_0) \Delta t \approx \int_a^b F(C(t)) \cdot C'(t) \, dt$$

**Physical Understanding #2: Vector Line Integrals = Circulation of Fluid**

Let  $F$  measure the flow of a fluid

$$\int_a^b F(C(t)) \cdot C'(t) \, dt = \int_a^b F(C(t)) \cdot \frac{C'(t)}{\|C'(t)\|} \|C'(t)\| \, dt$$

- $F(C(t)) \cdot \frac{C'(t)}{\|C'(t)\|} = \|F(C(t))\| \frac{\|C'(t)\|}{\|C'(t)\|} \cos(\theta) = \|F(C(t))\| \cos(\theta)$  is the component of  $F$  that moves in direction of  $C'(t)$  which is tangent to  $C(t)$
- $\|C'(t)\| \, dt$  is the change in arclength

Thus,  $\int_a^b F(C(t)) \cdot \frac{C'(t)}{\|C'(t)\|} \|C'(t)\| \, dt$  is the total change in fluid along  $C(t)$ .

Thus, the vector line integral is also called the **circulation of  $F$  along  $C(t)$** .

If  $F(x,y,z) = (M(x,y,z), N(x,y,z), P(x,y,z))$  and  $C(t) = (x(t), y(t), z(t))$ :

$$\int_{C(t)} F \cdot ds = \int_a^b M(x,y,z) \, dx + N(x,y,z) \, dy + P(x,y,z) \, dz$$

**Example**

Let  $F(x,y) = (x+y, x-y)$ . Find the vector line integral of  $F$  along  $x^2 + y^2 = 1$ .

$$C(t) = (\cos(t), \sin(t)) \text{ for } t \in [0, 2\pi].$$

$$\begin{aligned} \int_{C(t)} F \cdot ds &= \int_0^{2\pi} (\cos(t) + \sin(t), \cos(t) - \sin(t)) \cdot (-\sin(t), \cos(t)) \, dt \\ &= \int_0^{2\pi} -2\sin(t)\cos(t) + \cos^2(t) - \sin^2(t) \, dt \\ &= \int_0^{2\pi} -\sin(2t) + \cos(2t) \, dt = \left[ \frac{1}{2} \cos(2t) + \frac{1}{2} \sin(2t) \right]_0^{2\pi} = 0 \end{aligned}$$

**Definition 5.3.3: Reparametrization for Paths**

Let  $C_1: [a,b] \rightarrow \mathbb{R}^n$  be a piecewise  $C^1$  path.

Then piecewise  $C^1$  path  $C_2: [c,d] \rightarrow \mathbb{R}^n$  is a **reparametrization** of  $C_1$  if there is a 1-1 and onto  $v: [c,d] \rightarrow [a,b]$  such that:

$$C_2(t) = C_1(v(t))$$

Thus, since continuous  $v(t)$  is 1-1 and onto, then either:

- Orientation preserving:**  $v(c) = a$  and  $v(d) = b$  so  $v'(t) \geq 0$
- Orientation reversing:**  $v(c) = b$  and  $v(d) = a$  so  $v'(t) \leq 0$

**Theorem 5.3.4: Scalar Line Integrals are unaffected by Reparametrization**

Let  $C^1$  path  $C_1(t): [a,b] \rightarrow \mathbb{R}^n$  and  $f: C([a,b]) \subset X \subset \mathbb{R}^n \rightarrow \mathbb{R}$  be continuous.

Let  $C^1$  path  $C_2(t): [c,d] \rightarrow \mathbb{R}^n$  be a reparametrization of  $C_1(t)$ . Then:

$$\int_{C_2(t)} f \, ds = \int_{C_1(t)} f \, ds$$

**Proof**

$$\begin{aligned} \int_{C_2(t)} f \, ds &= \int_c^d f(C_2(t)) \|C_2'(t)\| \, dv = \int_c^d f(C_1(v(t))) \|C_1'(v(t))v'(t)\| \, dt \\ &= \int_c^d f(C_1(v(t))) \|C_1'(v(t))\| |v'(t)| \, dt \end{aligned}$$

If  $v(t)$  is orientation preserving, then:

$$\begin{aligned} \int_{C_2(t)} f \, ds &= \int_c^d f(C_1(v(t))) \|C_1'(v(t))\| v'(t) \, dt = \int_a^b f(C_1(v)) \|C_1'(v)\| \, dv \\ &= \int_{C_1(t)} f \, ds \end{aligned}$$

If  $v(t)$  is orientation reversing, then:

$$\begin{aligned} \int_{C_2(t)} f \, ds &= \int_c^d f(C_1(v(t))) \|C_1'(v(t))\| [-v'(t)] \, dt = - \int_b^a f(C_1(v)) \|C_1'(v)\| \, dv \\ &= \int_a^b f(C_1(v)) \|C_1'(v)\| \, dv = \int_{C_1(t)} f \, ds \end{aligned}$$

**Theorem 5.3.5: Vector Line Integrals are affected by Reparametrization**

Let  $C^1$  path  $C_1(t): [a,b] \rightarrow \mathbb{R}^n$  and  $F: C([a,b]) \subset X \subset \mathbb{R}^n \rightarrow \mathbb{R}^n$  be continuous.

Let  $C^1$  path  $C_2(t): [c,d] \rightarrow \mathbb{R}^n$  be a reparametrization of  $C_1(t)$ . Then:

- If  $C_2(t)$  is orientation preserving:

$$\int_{C_2(t)} F \cdot ds = \int_{C_1(t)} F \cdot ds$$

- If  $C_2(t)$  is orientation reversing:

$$\int_{C_2(t)} F \cdot ds = - \int_{C_1(t)} F \cdot ds$$

**Proof**

$$\int_{C_2(t)} F \cdot ds = \int_c^d F(C_2(t)) \cdot C_2'(t) \, dt = \int_c^d F(C_1(v(t))) \cdot C_1'(v(t))v'(t) \, dt$$

If  $v(t)$  is orientation preserving, then:

$$\int_{C_2(t)} F \cdot ds = \int_a^b F(C_1(v)) \cdot C_1'(v) \, dv = \int_{C_1(t)} F \cdot ds$$

If  $v(t)$  is orientation reversing, then:

$$\int_{C_2(t)} F \cdot ds = \int_b^a F(C_1(v)) \cdot C_1'(v) \, dv = - \int_{C_1(t)} F \cdot ds$$

## 5.4 Green's Theorem

### Definition 5.4.1: Simple, Closed Curves

$C^1$  curve  $C(t): [a,b] \rightarrow \mathbb{R}^n$  is **simple** if  $C(t)$  have no self-intersections (i.e. 1-1) except possibly at endpoints.

$C(t)$  is **closed** if endpoints are the same (i.e.  $C(a) = C(b)$ ).

### Theorem 5.4.2: Green's Theorem

Let  $D$  be a closed, bounded region in  $\mathbb{R}^2$  where the boundary of  $D$ ,  $C$  consist of finitely many simple, closed, piecewise  $C^1$  curves  $C_i$ . For each  $C_i$ , let parametrization  $C_i(t)$  be such that as  $t$  increases,  $D$  is at the left of  $C_i(t)$ .

Then for  $C^1$  vector field  $F(x,y) = (M(x,y), N(x,y))$ :

$$\oint_C F \cdot ds = \int \int_D -\frac{\partial M}{\partial y} + \frac{\partial N}{\partial x} dA = \int \int_D (\nabla \times F) \cdot e_3 dA$$

### Proof

Note  $\oint_{C(t)} F \cdot ds = \oint M(x,y) dx + N(x,y) dy$ .

First, split  $C = C_1 + C_2$  such that  $C_1$  is the curve below and  $C_2$  is the curve above. Parametrize  $C_1(x): [a,b] \rightarrow \mathbb{R}^2$  and  $C_2(x): [a,b] \rightarrow \mathbb{R}^2$ .

$$\begin{aligned} \oint_C M(x,y) dx &= \int_{C_1} M(x,y) dx - \int_{C_2} M(x,y) dx \\ &= \int_a^b M(x, C_1(x)) dx - \int_a^b M(x, C_2(x)) dx \\ &= \int_a^b M(x, C_1(x)) - M(x, C_2(x)) dx \end{aligned}$$

Since  $\int \int_D -\frac{\partial M}{\partial y} dA = \int_a^b \int_{C_1(x)}^{C_2(x)} -\frac{\partial M}{\partial y} dy dx = \int_a^b -M(x, C_2(x)) + M(x, C_1(x)) dx$  then  $\oint_{C(t)} M(x,y) dx = \int \int_D -\frac{\partial M}{\partial y} dA$ .

Next, split  $C = C_3 + C_4$  such that  $C_3$  is the curve to the left and  $C_4$  is the curve to the right. Parametrize  $C_3(y): [c,d] \rightarrow \mathbb{R}^2$  and  $C_4(y): [c,d] \rightarrow \mathbb{R}^2$ .

$$\begin{aligned} \oint_C N(x,y) dy &= -\int_{C_3} N(x,y) dy + \int_{C_4} N(x,y) dy \\ &= -\int_c^d N(C_3(y), y) dy + \int_c^d N(C_4(y), y) dy \\ &= \int_c^d -N(C_3(y), y) + N(C_4(y), y) dy \end{aligned}$$

Since  $\int \int_D \frac{\partial N}{\partial x} dA = \int_c^d \int_{C_3(y)}^{C_4(y)} \frac{\partial N}{\partial x} dx dy = \int_c^d N(C_4(y), y) - N(C_3(y), y) dy$  then  $\oint_{C(t)} N(x,y) dy = \int \int_D \frac{\partial N}{\partial x} dA$ .

Thus,  $\oint_C F \cdot ds = \oint M(x,y) dx + N(x,y) dy = \int \int_D -\frac{\partial M}{\partial y} + \frac{\partial N}{\partial x} dA$  and since  $\nabla \times F = (0, 0, \frac{\partial N}{\partial x} - \frac{\partial M}{\partial y})$ , then  $\int \int_D -\frac{\partial M}{\partial y} + \frac{\partial N}{\partial x} dA = \int \int_D (\nabla \times F) \cdot e_3 dA$

### Example

Find the area of an ellipse  $\frac{x^2}{a^2} + \frac{y^2}{b^2} = 1$ .

The parametrization of an ellipse is  $C(t) = (a \cos(t), b \sin(t))$  for  $t \in [0, 2\pi]$ .

Let  $F(x,y) = \frac{-y}{2}, \frac{x}{2}$ . Thus:

$$\begin{aligned} \text{Area} &= \int \int_{\frac{x^2}{a^2} + \frac{y^2}{b^2} = 1} 1 dA = \int \int_{\frac{x^2}{a^2} + \frac{y^2}{b^2} = 1} -\frac{\partial (-y)}{\partial y} + \frac{\partial x}{\partial x} dA = \int_C F \cdot ds \\ \int_C F \cdot ds &= \int_0^{2\pi} \left(-\frac{b}{2} \sin(t), \frac{a}{2} \cos(t)\right) \cdot (-a \sin(t), b \cos(t)) dt \\ &= \int_0^{2\pi} \frac{ab}{2} \sin^2(t) + \frac{ab}{2} \cos^2(t) dt = \int_0^{2\pi} \frac{ab}{2} dt = ab\pi \end{aligned}$$

**Theorem 5.4.3: Gauss's Theorem: Divergence Theorem in  $\mathbb{R}^2$** 

If Green's Theorem applies for region  $D$  where  $C$  is the boundary of  $D$  and  $\mathbf{n}$  is an outward unit normal vector to  $D$ , then for  $C^1$  vector field

$$\mathbf{F}(x,y) = (M(x,y), N(x,y)):$$

$$\oint_C \mathbf{F} \cdot \mathbf{n} \, ds = \iint_D \nabla \cdot \mathbf{F} \, dA$$

**Proof**

Let  $C$  be parametrized by  $C(t) = (x(t), y(t))$  for  $t \in [a, b]$ . Then the outward unit normal vector at  $t$  is  $\mathbf{n} = \frac{1}{\sqrt{(y'(t))^2 + (-x'(t))^2}}(y'(t), -x'(t)) = \frac{1}{\|C'(t)\|}(y'(t), -x'(t))$ .

$$\begin{aligned} \oint_C \mathbf{F} \cdot \mathbf{n} \, ds &= \int_a^b (M(x(t), y(t)), N(x(t), y(t))) \cdot \frac{1}{\|C'(t)\|}(y'(t), -x'(t)) \|C'(t)\| \, dt \\ &= \int_a^b M(x(t), y(t))y'(t) \, dt - \int_a^b N(x(t), y(t))x'(t) \, dt \\ &= \int_C M(x, y) \, dy - \int_C N(x, y) \, dx = \int_C (-N, M) \cdot ds \end{aligned}$$

By **theorem 5.4.2**:

$$\int_C (-N, M) \cdot ds = \iint_D \frac{\partial N}{\partial y} + \frac{\partial M}{\partial x} \, dA = \iint_D \nabla \cdot \mathbf{F} \, dA$$

**Example**

Let  $\mathbf{F}(x,y) = (-y, x)$  and region  $D: x^2 + y^2 \leq 1$ . Find  $\iint_D \text{div}(\mathbf{F}) \, dA$ .

Since  $\iint_D \text{div}(\mathbf{F}) \, dA = \iint_D \nabla \cdot \mathbf{F} \, dA$  where the boundary of  $D$ ,  $C(t) = (\cos(t), \sin(t))$  for  $t \in [0, 2\pi]$  so  $C'(t) = (-\sin(t), \cos(t))$  and  $\mathbf{n} = (\cos(t), \sin(t))$ . Thus:

$$\begin{aligned} \iint_D \text{div}(\mathbf{F}) \, dA &= \oint_C \mathbf{F} \cdot \mathbf{n} \, ds \\ &= \int_0^{2\pi} (-\sin(t), \cos(t)) \cdot (\cos(t), \sin(t)) \sqrt{(-\sin(t))^2 + (\cos(t))^2} \, dt \\ &= \int_0^{2\pi} 0 \, dt = 0 \end{aligned}$$

**5.5 Conservative Vector Fields****Definition 5.5.1: Path Independence**

A continuous vector field  $\mathbf{F}$  has **path independent** line integrals if for any two simple, piecewise  $C^1$ , oriented curves  $C_1, C_2$  on  $[a, b]$  where  $C_1(a) = C_2(a)$  and  $C_1(b) = C_2(b)$ :

$$\int_{C_1} \mathbf{F} \cdot ds = \int_{C_2} \mathbf{F} \cdot ds$$

**Theorem 5.5.2: Path Independence  $\Leftrightarrow \int_{C_1} \mathbf{F} \cdot ds = 0$** 

Continuous vector field  $\mathbf{F}$  has path independent line integrals if and only if

$$\oint_C \mathbf{F} \cdot ds = 0 \text{ for all simple, piecewise } C^1, \text{ closed curves.}$$

**Proof**

Suppose  $\mathbf{F}$  has path independence.

For simple, piecewise  $C^1$ , closed curve  $C$ , split  $C$  into  $C_1(t), C_2(t)$  for  $t \in [a, b]$  such that  $C_1(a) = C_2(b)$  and  $C_1(b) = C_2(a)$ . Since  $\mathbf{F}$  has path independence:

$$\oint_{C(t)} \mathbf{F} \cdot ds = \int_{C_1(t)} \mathbf{F} \cdot ds - \int_{C_2(t)} \mathbf{F} \cdot ds = 0$$

Suppose  $\oint_C \mathbf{F} \cdot ds = 0$  for all simple, piecewise  $C^1$ , closed curves  $C$ .

Then for simple, piecewise  $C^1$  curves  $C_1(t), C_2(t)$  for  $t \in [a, b]$  such that  $C_1(a) = C_2(a)$  and  $C_1(b) = C_2(b)$ , a simple, piecewise  $C^1$ , closed curve  $C = C_1 - C_2$ . Thus:

$$0 = \oint_C \mathbf{F} \cdot ds = \int_{C_1} \mathbf{F} \cdot ds - \int_{C_2} \mathbf{F} \cdot ds \Rightarrow \int_{C_1} \mathbf{F} \cdot ds = \int_{C_2} \mathbf{F} \cdot ds$$

**Definition 5.5.3: Potential of F**

Let continuous vector field  $F: \mathbb{R}^n \rightarrow \mathbb{R}^n$  have a  $C^1$  function  $f: \mathbb{R}^n \rightarrow \mathbb{R}$  such that  $\nabla f = F$ . Then  $f$  is the **potential** of  $F$  and  $F$  is a **conservative vector field**.

**Theorem 5.5.4: Fundamental Theorem of Line Integrals**

Let continuous vector field  $F: \mathbb{R}^n \rightarrow \mathbb{R}^n$ . Then  $F = \nabla f$  if and only if  $F$  has path independent line integrals.

If  $F = \nabla f$  for some  $C^1$  function  $f$ , then for any piecewise  $C^1$ , oriented curve  $C$  with initial point  $A$  and terminal point  $B$ :

$$\int_C F \cdot ds = f(B) - f(A)$$

**Proof**

Suppose  $F = \nabla f$  for some  $C^1$  function  $f$ .

Then  $Df(x)$  exist so  $\frac{d}{dt}f(C(t)) = Df(C(t)) \cdot C'(t) = \nabla f(C(t)) \cdot C'(t)$  for any piecewise  $C^1$ , oriented curve  $C$ . Let  $C$  have initial point  $(a,A)$  and terminal point  $(b,B)$ . Then:

$$\int_C F \cdot ds = \int_a^b \nabla f(C(t)) \cdot C'(t) dt = \int_a^b \frac{d}{dt}f(C(t)) dt = [f(C(t))]_a^b = f(B) - f(A)$$

Since  $\int_C F \cdot ds = f(B) - f(A)$  regardless of any piecewise  $C^1$ , oriented curve  $C(t)$ , then  $F$  has path independence.

Suppose  $F$  has path independence.

Take  $n = 2$ . The proof is analogous for any  $n$ .

For  $F = (F_1, F_2)$ , let  $f(x,y) = \int_C F \cdot ds = \int_{(0,0)}^{(x,y)} F_1 dx + F_2 dy$  where  $C$  is any piecewise  $C^1$  curve from  $(0,0)$  to  $(x,y)$ . Since  $F$  is path independent, then:

$$\begin{aligned} \int_{(0,0)}^{(x,y)} F_1 dx + F_2 dy &= [\int_{(0,0)}^{(0,y)} F_1 dx + F_2 dy] + [\int_{(0,y)}^{(x,y)} F_1 dx + F_2 dy] \\ &= \int_0^y F(0,t) \cdot (0,1) dt + \int_0^x F(t,y) \cdot (1,0) dt = \int_0^y F_2(0,t) dt + \int_0^x F_1(t,y) dt \end{aligned}$$

By the Fundamental Theorem of Calculus,  $\frac{\partial f}{\partial x} = 0 + F_1(x,y) = F_1(x,y)$ . Also:

$$\begin{aligned} \int_{(0,0)}^{(x,y)} F_1 dx + F_2 dy &= [\int_{(0,0)}^{(x,0)} F_1 dx + F_2 dy] + [\int_{(x,0)}^{(x,y)} F_1 dx + F_2 dy] \\ &= \int_0^x F(t,0) \cdot (1,0) dt + \int_0^y F(x,t) \cdot (0,1) dt = \int_0^x F_1(t,0) dt + \int_0^y F_2(x,t) dt \end{aligned}$$

By the Fundamental Theorem of Calculus,  $\frac{\partial f}{\partial y} = 0 + F_2(x,y) = F_2(x,y)$ .

Thus,  $\nabla f = (F_1, F_2) = F$ .

**Example**

Recall the example at 5.3.2. Let  $F(x,y) = (x+y, x-y)$ .

Find the vector line integral of  $F$  along  $x^2 + y^2 = 1$  from  $(1,0)$  to  $(0,1)$ .

Also, justify why  $\int_C F \cdot ds = 0$  over the unit circle.

Since  $F = \nabla f$  where  $f = \frac{1}{2}x^2 + xy - \frac{1}{2}y^2$ , then:

$$\int_C F \cdot ds = f(0,1) - f(1,0) = -\frac{1}{2} - \frac{1}{2} = -1$$

Additionally, the vector line integral of  $F$  along the entire  $x^2 + y^2 = 1$  which is closed so initial point  $A$  is equal to terminal point  $B$ :

$$\int_C F \cdot ds = f(B) - f(A) = f(A) - f(A) = 0$$

which is exactly the value calculated in the previous example.

## 6 Surface Integrals

### 6.1 Parametrized Surfaces

#### Definition 6.1.1: Parametrized Surfaces

A **parametrized surface** is a continuous function  $X: D \subset \mathbb{R}^2 \rightarrow \mathbb{R}^3$ .

Let surface  $S$  be parametrized by differentiable  $X(s,t) = (x(s,t), y(s,t), z(s,t))$ .

Then the tangent vectors:

$$T_s(s,t) = \frac{\partial X}{\partial s}(s,t) = \left( \frac{\partial x}{\partial s}(s,t), \frac{\partial y}{\partial s}(s,t), \frac{\partial z}{\partial s}(s,t) \right)$$

$$T_t(s,t) = \frac{\partial X}{\partial t}(s,t) = \left( \frac{\partial x}{\partial t}(s,t), \frac{\partial y}{\partial t}(s,t), \frac{\partial z}{\partial t}(s,t) \right)$$

$S$  is **smooth** if  $X$  is  $C^1$  and the **standard normal vector** of  $X$ :

$$N_X(s,t) = T_s(s,t) \times T_t(s,t) \neq 0.$$

$S$  is **piecewise smooth** if  $X = \{X_1, \dots, X_m\}$  where each parametrized surface  $X_i$  is  $C^1$  and  $N_{X_i}(s,t) \neq 0$

#### Theorem 6.1.2: Surface Area

Let surface  $S$  be parametrized by  $C^1$  function  $X: D \subset \mathbb{R}^2 \rightarrow \mathbb{R}^3$ .

Then the surface area of  $S$ :

$$\iint_D \|N_X(s,t)\| \, dA = \iint_D \sqrt{\left(\frac{\partial(x,y)}{\partial(s,t)}\right)^2 + \left(\frac{\partial(x,z)}{\partial(s,t)}\right)^2 + \left(\frac{\partial(y,z)}{\partial(s,t)}\right)^2} \, dA$$

If the surface  $S$  is  $z = f(x,y)$ , then the surface area of  $S$ :

$$\iint_D \sqrt{f_x^2 + f_y^2 + 1} \, dA$$

#### Proof

Let  $P = \{s_0, \dots, s_n\} \times \{t_0, \dots, t_n\}$  be a partition of region  $D$ . For each  $\Delta s_i \times \Delta t_j$ , choose a point  $(s_i^*, t_j^*) \in [s_{i-1}, s_i] \times [t_{j-1}, t_j]$ . Let  $S$  be the set of all points  $(s_i^*, t_j^*)$ . Then for each  $\Delta s_i \times \Delta t_j$ , the  $\|T_s(s_i^*, t_j^*)\Delta s_i \times T_t(s_i^*, t_j^*)\Delta t_j\|$  is an approximate area for the surface at  $\Delta s_i \times \Delta t_j \in D$ . Thus:

$$\begin{aligned} \text{Surface Area} &= \lim_{\|P\| \rightarrow 0} \sum_{i,j=1}^n \|T_s(s_i^*, t_j^*)\Delta s_i \times T_t(s_i^*, t_j^*)\Delta t_j\| \\ &= \lim_{\|P\| \rightarrow 0} \sum_{i,j=1}^n \|T_s(s_i^*, t_j^*) \times T_t(s_i^*, t_j^*)\| \Delta s_i \Delta t_j = \iint_D \|N_X(s,t)\| \, dA \end{aligned}$$

#### Example

Prove that the surface area of a sphere of radius  $R$  is  $4\pi R^2$ .

Using spherical coordinates,  $x = R \sin(\phi) \cos(\theta)$ ,  $y = R \sin(\phi) \sin(\theta)$ , and  $z = R \cos(\phi)$  for  $\phi \in [0, \pi]$  and  $\theta \in [0, 2\pi]$ . Let  $X(\theta, \phi)$  be a parametrization of the sphere. Then:

$$\begin{aligned} N_X(\theta, \phi) &= \begin{vmatrix} e_1 & e_2 & e_3 \\ -R \sin(\phi) \sin(\theta) & R \sin(\phi) \cos(\theta) & 0 \\ R \cos(\phi) \cos(\theta) & R \cos(\phi) \sin(\theta) & -R \sin(\phi) \end{vmatrix} \\ &= (-R^2 \sin^2(\phi) \cos(\theta), -R^2 \sin^2(\phi) \sin(\theta), -R^2 \sin(\phi) \cos(\phi)) \\ \int_0^{2\pi} \int_0^\pi \|(-R^2 \sin^2(\phi) \cos(\theta), -R^2 \sin^2(\phi) \sin(\theta), -R^2 \sin(\phi) \cos(\phi))\| \, d\phi d\theta \\ &= \int_0^{2\pi} \int_0^\pi R^2 \sin(\phi) \, d\phi d\theta = \int_0^{2\pi} 2R^2 \, d\theta = 4\pi R^2 \end{aligned}$$



## 6.2 Surface Integrals

### Definition 6.2.1: Scalar Surface Integrals

Let surface  $S$  be parametrized by smooth  $X(s,t): D \subset \mathbb{R}^2 \rightarrow \mathbb{R}^3$  and

$f(x,y,z): S \subset \mathbb{R} \subset \mathbb{R}^3 \rightarrow \mathbb{R}$  be continuous.

Then the **scalar surface integral** of  $f$  along  $X(s,t)$ :

$$\int \int_{X(s,t)} f \, dS = \int \int_D f(X(s,t)) \|N_X(s,t)\| \, dA$$

### Example

Let  $f(x,y,z) = 6xy$ . Find the scalar surface integral of  $f$  along the surface  $x+y+z=1$  in the first octant.

Let  $X(s,t) = (s,t,1-s-t)$  for  $s \in [0,1]$  and  $t \in [0,1-s]$ .

$$\begin{aligned} X_s(s,t) &= (1,0,-1) & X_t(s,t) &= (0,1,-1) & N_X(s,t) &= (1,1,1) \\ \int \int_{X(s,t)} f \, dS &= \int_0^1 \int_0^{1-s} 6st \sqrt{1^2 + 1^2 + 1^2} \, dt ds = 3\sqrt{3} \int_0^1 s - 2s^2 + s^3 \, ds = \frac{\sqrt{3}}{4} \end{aligned}$$

### Definition 6.2.2: Vector Surface Integrals

Let surface  $S$  be parametrized by smooth  $X(s,t): D \subset \mathbb{R}^2 \rightarrow \mathbb{R}^3$  and vector field  $F(x,y,z): S \subset \mathbb{R} \subset \mathbb{R}^3 \rightarrow \mathbb{R}^3$  be continuous.

Then the **vector surface integral** of  $F$  along  $X(s,t)$ :

$$\int \int_{X(s,t)} F \cdot dS = \int \int_D F(X(s,t)) \cdot N_X(s,t) \, dA$$

### Physical Understanding: Vector Surface Integrals = Divergence of Fluid

Let  $F$  measure the flow of water.

$$\int \int_D F(X(s,t)) \cdot N_X(s,t) \, dA = \int \int_D F(X(s,t)) \cdot \frac{N_X(s,t)}{\|N_X(s,t)\|} \|N_X(s,t)\| \, dA$$

- $F(X(s,t)) \cdot \frac{N_X(s,t)}{\|N_X(s,t)\|} = \|F(X(s,t))\| \frac{\|N_X(s,t)\|}{\|N_X(s,t)\|} \cos(\theta) = \|F(X(s,t))\| \cos(\theta)$   
is the component of  $F$  that moves in direction of  $N_X(s,t)$  which is orthogonal to  $F(X(s,t))$ .
- $\|N_X(s,t)\| \, dA$  is the change in surface area

Thus,  $\int \int_D F(X(s,t)) \cdot \frac{N_X(s,t)}{\|N_X(s,t)\|} \|N_X(s,t)\| \, dA$  is the total change in fluid across the surface  $X(s,t)$ . Thus, the vector surface integral is also called the **flux of  $F$  across  $S = X(s,t)$** .

### Example

Let  $F(x,y,z) = (x^3, -y^3, 0)$  and surface  $S$  be bounded by  $x^2 + y^2 = 4$  from  $z \in [0,4]$  with circles at  $z=0, z=4$ . Let the normal vectors be pointing outwards.

Find the vector surface integral of  $F$  along surface  $S$ .

Surface  $z$  consist of 3 parts:

$$S_1: x^2 + y^2 = 4, z=0 \quad S_2: x^2 + y^2 = 4, z=4 \quad S_3: x^2 + y^2 = 4, z \in [0,4]$$

Since the outward unit normal vector for  $S_1 = (0,0,1)$  and  $S_2 = (0,0,-1)$ , then:

$$\begin{aligned} \int \int_S F \cdot dS &= \iint_{D_1} F(X_{S_1}) \cdot n_1 \, dA + \iint_{D_2} F(X_{S_2}) \cdot n_2 \, dA + \iint_{D_3} F(X_{S_3}) \cdot n_3 \, dA \\ &= 0 + 0 + \int \int_{D_3} F(X_{S_3}) \cdot n_3 \, dA = \int \int_{D_3} F(X_{S_3}) \cdot n_3 \, dA \end{aligned}$$

$S_3$  can be parametrized  $X(s,t) = (2 \cos(s), 2 \sin(s), t)$  where  $s \in [0, 2\pi]$ ,  $t \in [0, 4]$ .

$$N_X(s,t) = (-2 \sin(s), 2 \cos(s), 0) \times (0, 0, 1) = (2 \cos(s), 2 \sin(s), 0)$$

which has the outward orientation so  $N_X(s,t)$  is correct.

$$\begin{aligned} \int \int_{D_3} F(X_{S_3}) \cdot n_3 \, dA &= \int_0^{2\pi} \int_0^4 16 \cos^4(s) - 16 \sin^4(s) \, dt ds \\ &= 64 \int_0^{2\pi} \cos(2s) \, ds = 0 \end{aligned}$$

**Definition 6.2.3: Reparametrization for Surfaces**

Let  $X_1: D_1 \subset \mathbb{R}^2 \rightarrow \mathbb{R}^3$  be a smooth parametrization of surface  $S$ .

Then smooth  $X_2: D_2 \subset \mathbb{R}^2 \rightarrow \mathbb{R}^3$  is a **reparametrization** of  $X_1$  if there is a 1-1 and onto  $v: D_2 \rightarrow D_1$  such that:

$$X_2(s, t) = X_1(v(s, t))$$

Let  $X_1(v_1, v_2) = (x_1, y_1, z_1)$ ,  $X_2(s, t) = (x_2, y_2, z_2)$ , and  $v(s, t) = (v_1, v_2)$ . Then:

$$\begin{bmatrix} \frac{\partial x_2}{\partial s} & \frac{\partial x_2}{\partial t} \\ \frac{\partial y_2}{\partial s} & \frac{\partial y_2}{\partial t} \\ \frac{\partial z_2}{\partial s} & \frac{\partial z_2}{\partial t} \end{bmatrix} = \begin{bmatrix} \frac{\partial x_1}{\partial v_1} & \frac{\partial x_1}{\partial v_2} \\ \frac{\partial y_1}{\partial v_1} & \frac{\partial y_1}{\partial v_2} \\ \frac{\partial z_1}{\partial v_1} & \frac{\partial z_1}{\partial v_2} \end{bmatrix} \begin{bmatrix} \frac{\partial v_1}{\partial s} & \frac{\partial v_1}{\partial t} \\ \frac{\partial v_2}{\partial s} & \frac{\partial v_2}{\partial t} \end{bmatrix} = \begin{bmatrix} \frac{\partial x_1}{\partial v_1} \frac{\partial v_1}{\partial s} + \frac{\partial x_1}{\partial v_2} \frac{\partial v_2}{\partial s} & \frac{\partial x_1}{\partial v_1} \frac{\partial v_1}{\partial t} + \frac{\partial x_1}{\partial v_2} \frac{\partial v_2}{\partial t} \\ \frac{\partial y_1}{\partial v_1} \frac{\partial v_1}{\partial s} + \frac{\partial y_1}{\partial v_2} \frac{\partial v_2}{\partial s} & \frac{\partial y_1}{\partial v_1} \frac{\partial v_1}{\partial t} + \frac{\partial y_1}{\partial v_2} \frac{\partial v_2}{\partial t} \\ \frac{\partial z_1}{\partial v_1} \frac{\partial v_1}{\partial s} + \frac{\partial z_1}{\partial v_2} \frac{\partial v_2}{\partial s} & \frac{\partial z_1}{\partial v_1} \frac{\partial v_1}{\partial t} + \frac{\partial z_1}{\partial v_2} \frac{\partial v_2}{\partial t} \end{bmatrix}$$

$$X_{2s} = X_{1v_1} \frac{\partial v_1}{\partial s} + X_{1v_2} \frac{\partial v_2}{\partial s}$$

$$X_{2t} = X_{1v_1} \frac{\partial v_1}{\partial t} + X_{1v_2} \frac{\partial v_2}{\partial t}$$

$$N_{X_2} = X_{2s} \times X_{2t} = [X_{1v_1} \times X_{1v_2}] \left( \frac{\partial v_1}{\partial s} \frac{\partial v_2}{\partial t} - \frac{\partial v_2}{\partial s} \frac{\partial v_1}{\partial t} \right) = N_{X_1} \frac{\partial(v_1, v_2)}{\partial(s, t)}$$

Thus:

**Orientation preserving:**  $\frac{\partial(v_1, v_2)}{\partial(s, t)} > 0$

**Orientation reversing:**  $\frac{\partial(v_1, v_2)}{\partial(s, t)} < 0$

**Theorem 6.2.4: Scalar Surface Integrals are unaffected by Reparametrization**

Let surface  $S$  be parametrized by smooth  $X_1(v_1, v_2): D_1 \subset \mathbb{R}^2 \rightarrow \mathbb{R}^3$  and  $f(x, y, z): S \subset \mathbb{R}^3 \rightarrow \mathbb{R}$  be continuous.

Let  $X_2(s, t): D_2 \rightarrow \mathbb{R}^3$  be a reparametrization of  $X_1(s, t)$ . Then:

$$\int \int_{X_2(s, t)} f \, dS = \int \int_{X_1(s, t)} f \, dS$$

**Proof**

$$\begin{aligned} \int \int_{X_2(s, t)} f \, dS &= \int \int_{D_2} f(X_2(s, t)) \|N_{X_2(s, t)}\| \, dA \\ &= \int \int_{D_2} f(X_1(v(s, t))) \|N_{X_1(v_1(s, t), v_2(s, t))} \frac{\partial(v_1, v_2)}{\partial(s, t)}\| \, dA \\ &= \int \int_{D_2} f(X_1(v(s, t))) \|N_{X_1(v_1(s, t), v_2(s, t))}\| \left| \frac{\partial(v_1, v_2)}{\partial(s, t)} \right| \, dA \end{aligned}$$

Then by **theorem 4.3.2**:

$$= \int \int_{D_1} f(X_1(v_1, v_2)) \|N_{X_1(v_1, v_2)}\| \, dA = \int \int_{X_1(s, t)} f \, dS$$

**Theorem 6.2.5: Vector Surface Integrals are affected by Reparametrization**

Let surface  $S$  be parametrized by smooth  $X_1(v_1, v_2): D_1 \subset \mathbb{R}^2 \rightarrow \mathbb{R}^3$  and vector field  $F(x, y, z): S \subset \mathbb{R}^3 \rightarrow \mathbb{R}^3$  be continuous.

Let  $X_2(s, t): D_2 \rightarrow \mathbb{R}^3$  be a reparametrization of  $X_1(s, t)$ . Then:

- If  $X_2(s, t)$  is orientation preserving:

$$\int \int_{X_2(s, t)} F \cdot dS = \int \int_{X_1(s, t)} F \cdot dS$$

- If  $X_2(s, t)$  is orientation reversing:

$$\int \int_{X_2(s, t)} F \cdot dS = - \int \int_{X_1(s, t)} F \cdot dS$$

**Proof**

$$\begin{aligned} \int \int_{X_2(s, t)} F \cdot dS &= \int \int_{D_2} F(X_2(s, t)) \cdot N_{X_2(s, t)} \, dA \\ &= \int \int_{D_2} F(X_1(v(s, t))) \cdot N_{X_1(v_1(s, t), v_2(s, t))} \frac{\partial(v_1, v_2)}{\partial(s, t)} \, dA \end{aligned}$$

If  $X_2(s, t)$  is orientation preserving, then by **theorem 4.3.2**:

$$= \int \int_{D_1} F(X_1(v_1, v_2)) \cdot N_{X_1(v_1, v_2)} \, dA = \int \int_{X_1(s, t)} F \cdot dS$$

If  $X_2(s, t)$  is orientation reversing, then by **theorem 4.3.2**:

$$= - \int \int_{D_1} F(X_1(v_1, v_2)) \cdot N_{X_1(v_1, v_2)} \, dA = - \int \int_{X_1(s, t)} F \cdot dS$$

## 6.3 Stokes's & Gauss's Theorems

### Definition 6.3.1: Induced Orientation

Let  $S$  be a bounded, piecewise smooth surface in  $\mathbb{R}^3$  with boundary defined by a simple, closed curve  $C$ . Let unit normal vector  $\mathbf{n}$  be orthogonal to  $S$  at a point so surface  $S$  has an orientation of  $\mathbf{n}$ .

Then for points on  $S$ , let there be a spinning disc with a direction such that the curl is in the same direction as  $\mathbf{n}$  at such points.

Then  $C$  has an **induced orientation** from  $S$  if the orientation of curve  $C$  is in the same direction as the spinning discs of points near  $C$ .

### Theorem 6.3.2: Stokes's Theorem

Let  $S$  be a bounded, piecewise smooth, oriented surface in  $\mathbb{R}^3$  with boundary defined by finitely many simple, closed, piecewise  $C^1$  curves  $C = \{C_1, \dots, C_m\}$  which have an induced orientation from  $S$ .

Then for  $C^1$  vector field  $F(x,y,z): S \subset \mathbb{R} \subset \mathbb{R}^3 \rightarrow \mathbb{R}^3$ :

$$\iint_S \nabla \times F \cdot d\mathbf{S} = \oint_C F \cdot d\mathbf{s}$$

### Example

Let  $F(x,y) = (x+y, x-y)$  and  $S$  be the hemisphere  $x^2 + y^2 + z^2 = 1$  for  $z \geq 0$ .

Prove  $\iint_S \nabla \times F \cdot d\mathbf{S} = 0$ .

Recall the example at 5.5.4 which used the example from 5.3.2.

Suppose  $S$  has outward normal vectors. The case for inward normal vectors is analogous. Since  $S$  is a bounded, piecewise smooth, oriented surface with a boundary  $C$ :  $x^2 + y^2 = 1$  with an induced orientation from  $S$ , then:

$$\iint_S \nabla \times F \cdot d\mathbf{S} = \oint_C F \cdot d\mathbf{s}$$

Recall the example at 5.5.4 and 5.3.2 calculated that the  $\oint_C F \cdot d\mathbf{s} = 0$ .

Thus,  $\iint_S \nabla \times F \cdot d\mathbf{S} = 0$ .

Note that region  $S$  is irrelevant, only its boundary is important.

Stokes' Theorem shows that given a  $C^1$  vector field, the vector surface integral of the curl of any surface is equivalent to the vector surface integral of the curl of another surface as long as the surfaces have the same boundary.

**Theorem 6.3.3: Gauss's Theorem: Divergence Theorem in  $\mathbb{R}^3$** 

Let  $D$  be a bounded region in  $\mathbb{R}^3$  with boundary defined by finitely many closed, piecewise smooth surfaces  $S = \{S_1, \dots, S_m\}$  oriented with unit normal vectors that points away from  $D$ .

Then for  $C^1$  vector field  $F(x,y,z): D \subset \mathbb{R} \subset \mathbb{R}^3 \rightarrow \mathbb{R}^3$ :

$$\oint_S F \cdot dS = \int \int \int_D \nabla \cdot F \, dV$$

**Example**

Recall the example at 6.2.2.

Let  $F(x,y,z) = (x^3, -y^3, 0)$  and surface  $S$  be bounded by  $x^2 + y^2 = 4$  from  $z \in [0,4]$  with circles at  $z=0, z=4$ . Let the normal vectors be pointing outwards.

Justify why  $\oint_S F \cdot dS = 0$ .

Since  $S$  is a boundary for bounded region  $D: x^2 + y^2 \leq 4$  from  $z \in [0,4]$  in which  $S$  is oriented with outward normal vectors, then:

$$\oint_S F \cdot dS = \int \int \int_D \nabla \cdot F \, dV$$

Region  $D: x^2 + y^2 \leq 4$  for  $z \in [0,4]$  be redefined in cylindrical coordinates such that  $r \in [0,2]$ ,  $\theta \in [0, 2\pi]$ , and  $z \in [0,4]$ .

$$\nabla \cdot F = \frac{\partial}{\partial x} x^3 + \frac{\partial}{\partial y} (-y^3) + \frac{\partial}{\partial z} 0 = 3x^2 - 3y^2$$

Applying [theorem 4.3.2](#) for  $x = r \cos(\theta)$ ,  $y = r \sin(\theta)$ , and  $z = z$ :

$$\frac{\partial(x,y,z)}{\partial(r,\theta,z)} = \begin{vmatrix} \cos(\theta) & -r \sin(\theta) & 0 \\ \sin(\theta) & r \cos(\theta) & 0 \\ 0 & 0 & 1 \end{vmatrix} = r$$

$$\begin{aligned} \int \int \int_D \nabla \cdot F \, dV &= \int \int \int_{x^2+y^2 \leq 4, z \in [0,4]} \nabla \cdot F \, dV \\ &= \int_0^4 \int_0^2 \int_0^{2\pi} [3 \cos^2(\theta) - 3 \sin^2(\theta)] r \, d\theta dr dz \\ &= \int_0^4 \int_0^2 \int_0^{2\pi} 3r \cos(2\theta) \, d\theta dr dz = \int_0^4 \int_0^2 0 \, dr dz = 0 \end{aligned}$$

which is exactly the value calculated in the previous example.

## References

- [1] Susan Jane Colley, *Vector Calculus (4th Edition)*, ISBN-13: 978-0321780652