

MTHE 280 - Lecture Notes

ADVANCED CALCULUS

Prof. Maria Teresa Chiri and Prof. Sunil Naik • Fall 2025 • Queen's University

Contents

| | | |
|----------|---|-----------|
| 1 | Introduction to Multivariable Functions | 3 |
| 1.1 | Properties of functions | 3 |
| 1.2 | Identify domain and codomain | 3 |
| 2 | Level Curves and Contours | 5 |
| 3 | Limits of a function | 6 |
| 3.1 | L'Hospital's Rule | 6 |
| 3.2 | Limits in two variables | 6 |
| 3.3 | Epsilon-delta definition of a limit | 6 |
| 3.3.1 | General solution process | 7 |
| 3.4 | When to use either strategy | 8 |
| 3.5 | $\varepsilon - \delta$ for vector-valued functions | 8 |
| 4 | Continuity and its properties | 9 |
| 4.1 | Continuity of single variable functions | 9 |
| 4.2 | Continuity of multivariable functions | 9 |
| 4.3 | Properties of continuity (scalar- and vector-valued functions) | 9 |
| 4.4 | Composition of two continuous functions | 10 |
| 5 | Differentiation of multivariable functions | 11 |
| 5.1 | The derivative | 11 |
| 5.2 | Notation | 11 |
| 5.3 | Partial Differentiation | 11 |
| 6 | Partial Differentiation (cont.) | 12 |
| 6.1 | Tangent plane visualized | 12 |
| 6.2 | Directional derivative | 12 |
| 6.3 | Multivariable differentiability at (a, b) | 12 |
| 7 | Gradients, More Derivatives, and the Jacobian | 14 |
| 7.1 | Gradient | 14 |
| 7.2 | Derivative Matrix | 14 |
| 7.3 | Differentiability in higher dimensions $f : U \rightarrow \mathbf{R}^m$ | 15 |
| 7.3.1 | Theorems for higher-dimension differentiability | 15 |

| | | |
|-----------|---|-----------|
| 7.4 | Properties of Differentiability | 15 |
| 8 | Differentiability in higher dimension | 17 |
| 8.1 | Chain Rule in Composition | 17 |
| 8.2 | Polar Coordinate Examples | 17 |
| 9 | Applications of the Gradient | 18 |
| 9.1 | Gradients and level curves | 18 |
| 9.2 | Magnitude of ∇F | 18 |
| 9.2.1 | Example | 18 |
| 10 | Conservative Vector Fields | 20 |
| 10.1 | Test for conservative | 20 |
| 10.2 | Reconstruct a potential function given its gradient | 20 |
| 11 | Parametric Equations and Class | 21 |
| 11.1 | Parametrization | 21 |
| 11.2 | Class | 21 |
| 12 | Arc Length, Divergence and Curl | 22 |
| 12.1 | Arc Length | 22 |
| 12.2 | Divergence of a vector field | 22 |
| 12.3 | Curl of a vector field | 22 |
| 13 | Gradient, Divergence and Curl (cont.) | 23 |
| 13.1 | Identities | 23 |
| 14 | Special Domains and Conservative Functions | 24 |
| 15 | Riemann Sums | 25 |
| 15.1 | Single-variable Integration | 25 |
| 15.2 | How to integrate functions of two variables | 25 |
| 16 | Cheat Sheet | 26 |
| 16.1 | Delta-Epsilon | 26 |
| 16.2 | Disproving a Multivariable Limit | 26 |
| 16.3 | Partial Derivative | 26 |
| 16.4 | Derivative Matrix | 26 |
| 16.5 | Divergence and Curl | 27 |

1 Introduction to Multivariable Functions

A function $f(x, y)$ is a rule that assigns to every element x a unique element y , and is denoted by $f : x \rightarrow y$, where x is the domain of f and y is the codomain of f

Example

$$f : \mathbf{N} \rightarrow \mathbf{R}, f(x) = 2x$$

In this case, every value of f is even and does not take the whole codomain

We introduce the range, a subset of the codomain, $range(f) \subseteq codomain(f)$

1.1 Properties of functions

One-one/Injective

$$f : X \rightarrow Y \text{ if } x_1, x_2 \in X, f(x_1) = f(x_2)$$

Onto/Surjective

$$f : X \rightarrow Y \text{ is onto if for every } y \in Y, \text{ there exists some } x \in X \text{ such that } f(x) = y$$

In this case, $codomain = range$

Bijjective

if $f : x \rightarrow y$ is both one-one and onto, it is bijective

Scalar-valued

Consider $f : x \rightarrow y$ where $x \subseteq \mathbf{R}$ and $y \subseteq \mathbf{R}$, $n, m \in \mathbf{N}$

When the codomain is just \mathbf{R} , the function is called a Scalar-valued function

Example

$$f : \mathbf{R}^2 \rightarrow \mathbf{R} \text{ where } f(x, y) = \sqrt{x^2 + y^2}$$

This returns the length of a 2D vector, which is a scalar

Vector-valued

A vector-valued function has codomain \mathbf{R}^n where $n > 1, n \in \mathbf{N}$

Example

$$f : \mathbf{R} \rightarrow \mathbf{R}^2, f(x) = (\cos x, \sin x)$$

1.2 Identify domain and codomain

Examples

$$f(x) = \ln x, \text{ domain} = (0, \infty), \text{ codomain} = \mathbf{R}$$

$$f(x) = \sqrt{2-x}, \text{ domain} = (-\infty, 2], \text{ codomain} = (0, \infty)$$

$$f(x, y) = (\sqrt{1-x^2-y^2}, \ln(y+1), x^2+y^2)$$

$$1: x^2 + y^2 = 1 \quad 2: y > -1$$

domain: $\{(x, y) \in \mathbf{R}^2 : x^2 + y^2 \leq 1, y > -1\}$

2 Level Curves and Contours

Level Curve

Given a scalar-valued function, the level curve at height c is the curve in \mathbf{R}^2 s.t. $f(x, y) = c$

Or, the level curve at height $c = \{(x, y) \in \mathbf{R}^2 | f(x, y) = c\}$

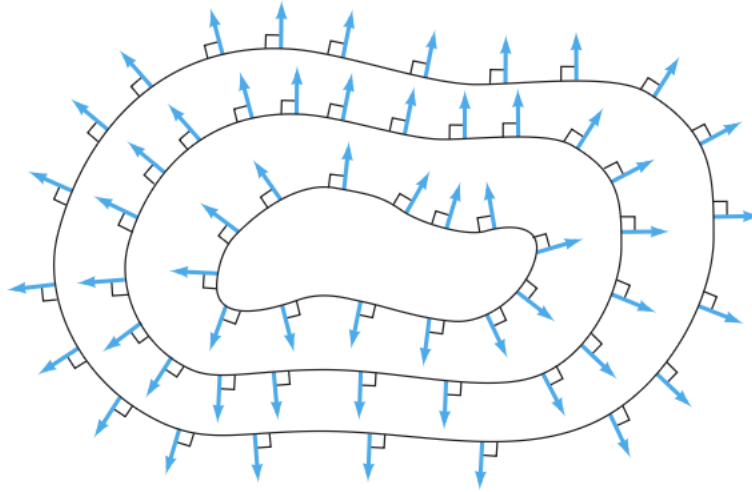


Figure 3.31 A gradient vector field $\mathbf{F} = \nabla f$. Equipotential lines are shown where f is constant.

Contour

The contour curve at height c is the collection of points (x, y, z) s.t. $z = f(x, y) = c$

Or, $\{(x, y, z) \in \mathbf{R}^3 | z = f(x, y) = c\}$

The projection of the contour is the level curve

Section

A section of a surface by a plane is just the intersection of the surface with that plane

3 Limits of a function

General form: $f : \mathbf{R} \rightarrow \mathbf{R}$

$\lim_{x \rightarrow a} f(x) = L, \therefore f(x)$ tends to L as x tends to a

3.1 L'Hospital's Rule

If we have a case where we are evaluating a limit and we get $\frac{0}{0}$ or $\frac{\infty}{\infty}$, we can use $\lim_{x \rightarrow a} \frac{f'(x)}{g'(x)}$

Why?: The ratio $\frac{f(x)}{g(x)}$ near a depends not only on the values of f and g , but on how fast they approach 0 or ∞

3.2 Limits in two variables

Let $f : \mathbf{R}^2 \rightarrow \mathbf{R}$, $\lim_{(x,y) \rightarrow (a,b)} f(x,y)$

The Line $y = mx$ trick

All paths approaching point (e.g. $(0,0)$) must give the same value

A simple test path is a straight line mx through the origin, and plug $f(x,y) \rightarrow f(x,mx)$

If the result depends on m , the limit does not exist

Does Exist Example

$$\lim_{(x,mx) \rightarrow (0,0)} \frac{x^2}{x^2 + y^4}$$

$$\lim_{x \rightarrow 0} \frac{x^2}{x^2 + m^4 x^4}$$

$$\lim_{x \rightarrow 0} \frac{1}{1 + m^4 x^2} = 1 \therefore \text{limit exists}$$

Does Not Exist Example

$$\lim_{(x,y) \rightarrow (0,0)} \frac{1}{1 + m^2} = \frac{x^2}{x^2 + m^2 x^2} = \frac{1}{1 + m^2} \therefore \text{limit does not exist}$$

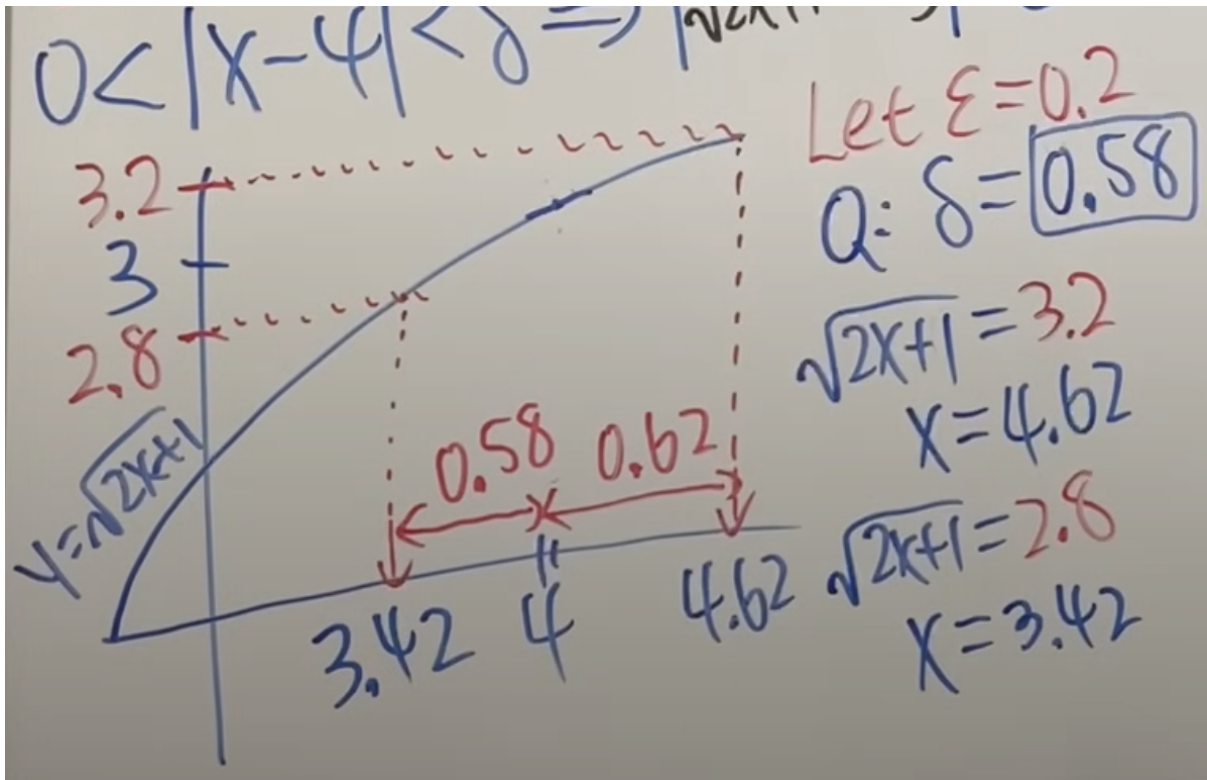
3.3 Epsilon-delta definition of a limit

$\lim_{x \rightarrow a} f(x) = L$ means $\forall \varepsilon > 0, \exists \delta > 0$ s.t. $0 < |x - a| < \delta \Rightarrow |f(x) - L| < \varepsilon$

Example: we know that $\lim_{x \rightarrow 4} \sqrt{2x+1} = 3$ by plugging in 4 into the continuous function

To prove this, $\forall \varepsilon > 0, \exists \delta > 0$ s.t. $0 < |x - 4| < \delta \Rightarrow |\sqrt{2x+1} - 3| < \varepsilon$

If x is near 4, of a distance less than δ , then the corresponding value of the function is near the limit $L = 3$, of a distance ε



3.3.1 General solution process

Proof: Given $\varepsilon > 0$ We want to find $\delta > 0$ such that if $0 < ||x - a|| < \delta$, then $|f(x) - L| < \varepsilon$

Start with $|f(x) - L|$ and manipulate it to relate it to $||x - a||$ For instance, show: $|f(x) - L| \leq c||x - a||$ for some $c > 0$

Choose $\delta = \frac{\varepsilon}{c}$ and show that $|f(x) - L| < c||x - a|| < c\delta = \varepsilon$

Therefore, $\lim_{x \rightarrow a} f(x) = L$

Triangle Inequality

It says: $|a + b| \leq |a| + |b|$

Order Trick

Ex: $\lim_{(x,y) \rightarrow (0,0)} \frac{3xy^2}{x^2+y^2} = 0$, lim is likely to exist when order is ≥ 1 , here it is 1

Simplify Trick

We can: $\frac{3|x|y^2}{x^2+y^2} \leq \frac{3|x|y^2}{y^2} = 3|x|$

We can also: $|x| \leq \sqrt{x^2 + y^2}$

Linear combination of coordinate differences

$$|a(x - a) + b(y - b)| \leq |a||x - a| + |b||y - b| \leq (|a| + |b|)||\mathbf{x} - \mathbf{a}||.$$

3.4 When to use either strategy

We use the epsilon-delta proof to rigorously prove that a limit exists (or equals some value)

We take the limit along lines, parabolas, or curves to test whether a limit exists, or to guess its value. It is useful when you are not sure if the limit exists.

3.5 $\varepsilon - \delta$ for vector-valued functions

Let $F : U(\subseteq \mathbf{R}^n) \rightarrow \mathbf{R}^m, \vec{a} \in U$

We write $\lim_{\vec{x} \rightarrow \vec{a}} F(\vec{x}) = \vec{L}, \forall \varepsilon > 0, \exists \delta > 0$ s.t. $\|F(\vec{x}) - \vec{L}\| < \varepsilon$ if $\|\vec{x} - \vec{a}\| < \delta$

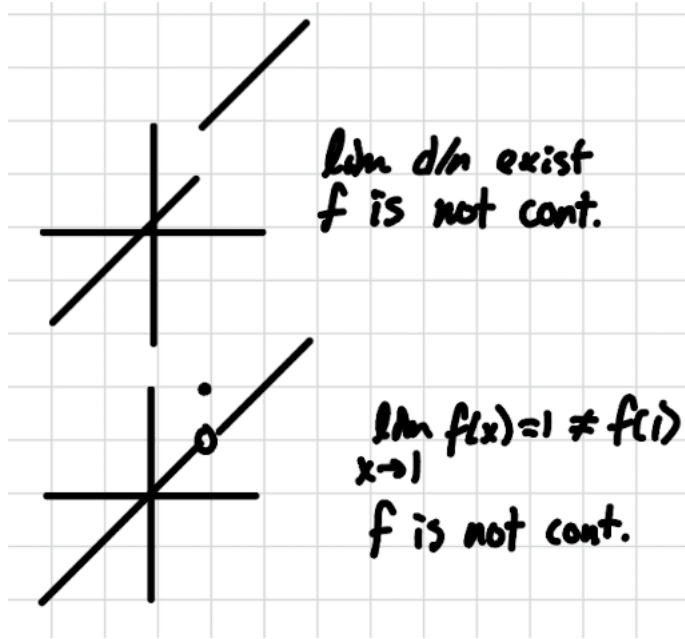
Ex: does $\lim_{(x,y) \rightarrow (0,0)} \left(\frac{3xy^2}{x^2+y^2}, \frac{e^x + \cos y}{x^2+y^2+1} \right)$ exist?

We know that the first component does. For the second component, both the numerator and the denominator are continuous at $(0,0)$, thus we can plug in that point and get that the limit approaches 2

4 Continuity and its properties

4.1 Continuity of single variable functions

Let $f : A \rightarrow \mathbb{R}, a \in A$. f is continuous if (1) $\lim_{x \rightarrow a} f(x)$ exists and (2) $\lim_{x \rightarrow a} f(x) = f(a)$



4.2 Continuity of multivariable functions

Let $f : U(\subseteq \mathbb{R}^n \rightarrow \mathbb{R}$ and $\vec{a} \in U$. f is continuous at \vec{a} if (1) $\lim_{\vec{x} \rightarrow \vec{a}} F(\vec{x})$ exists and (2) $\lim_{\vec{x} \rightarrow \vec{a}} F(\vec{x}) = F(\vec{a})$

4.3 Properties of continuity (scalar- and vector-valued functions)

Suppose that f and g are continuous at $\vec{a} \in U$

1. $f + g$ is continuous at \vec{a}
2. $f * g$ is continuous at \vec{a}
3. $\frac{f}{g}$ is continuous at \vec{a} if $g(\vec{a}) \neq 0$

Further:

1. $\lim_{\vec{x} \rightarrow \vec{a}} (f + g)(\vec{x}) = f(\vec{a}) + g(\vec{a})$
2. $\lim_{\vec{x} \rightarrow \vec{a}} (f * g)(\vec{x}) = f(\vec{a})g(\vec{a})$
3. $\lim_{\vec{x} \rightarrow \vec{a}} \left(\frac{f}{g} \right) (\vec{x}) = \frac{f(\vec{a})}{g(\vec{a})}$ if $g(\vec{a}) \neq 0$

Example:

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

For which values of a is F continuous?

We know that the first component is continuous everywhere, except possible at $(0, 0)$

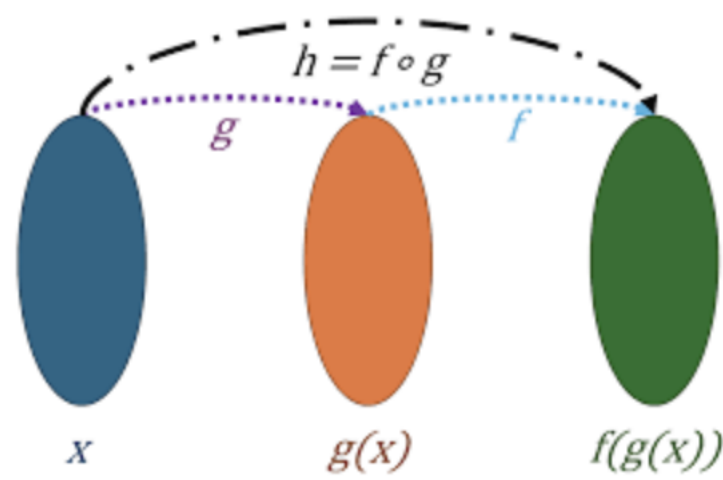
For continuity at $(0, 0)$, we need the limit of F at $(0, 0) = a$, which is equivalent to saying that the continuous function $F(0, 0) = a$

That means we need to compute the first term's limit while approaching $(0, 0)$, which is $= 0$

$\therefore a = 0$

4.4 Composition of two continuous functions

If: 1. g is continuous at $x = a$, and 2. f is continuous at $g(a)$, then $f \circ g$ is continuous at a , where $f(g(x)) \rightarrow f(g(a))$



5 Differentiation of multivariable functions

5.1 The derivative

f is differentiable at c if $\lim_{h \rightarrow c} \frac{f(x+h)-f(c)}{h}$ exists. If the limit exists, then it is denoted by $f'(x) = \lim_{h \rightarrow 0} \frac{f(x+h)-f(h)}{h}$, where $f'(x)$ captures the rate of change of f near c

If $f'(c)$ exists, we can draw a tangent line at c , and its slope is $f'(c)$

5.2 Notation

An **open ball** in \mathbf{R}^n with centre at $\vec{a} \in \mathbf{R}^n$ and radius $r : B(\vec{a}, r)$. The ball is open, meaning that the boundary points are not included

Definition: A point \vec{a} is an **interior point** of a set A if there exists an open ball $B_\varepsilon(\vec{a})$, for some $\varepsilon > 0$, such that $B_\varepsilon(\vec{a}) \subseteq A$. So, the open ball lies entirely inside the set, without touching its complement

Definition: A **boundary point** is a point \vec{a} such that every open ball $B_\varepsilon(\vec{a})$, no matter how small $\varepsilon > 0$ is, intersects the function and its complement (not the function)

Essentially, an open ball is all points strictly inside a certain radius from the centre, not including the edge. The interior points are inside the open ball, and boundary points are on the edge.

A set $U \subseteq \mathbf{R}^n$ is called open if every point of U is an interior point

5.3 Partial Differentiation

f is partially differentiable wrt x at (a, b) if $\lim_{x \rightarrow a} \frac{f(a+h, h) - f(a, h)}{h}$ exists. If exists: $\frac{\partial f}{\partial x}(a, b)$ or $f_x(a, b)$

6 Partial Differentiation (cont.)

6.1 Tangent plane visualized

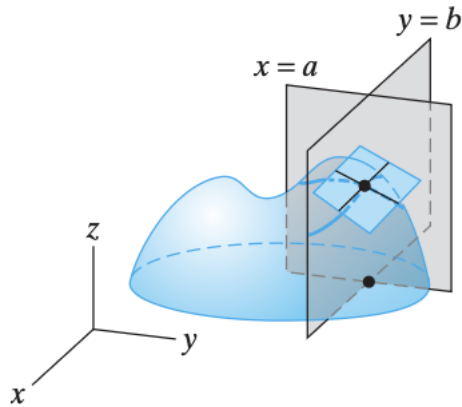


Figure 2.51 The **tangent plane** at $(a, b, f(a, b))$ contains the lines tangent to the curves formed by intersecting the surface $z = f(x, y)$ by the planes $x = a$ and $y = b$.

6.2 Directional derivative

The directional derivative of a function $f : \mathbb{R}^n \rightarrow \mathbb{R}$ at a point p in the direction of a vector \vec{v} is the rate at which f changes at p as you move in the direction of \vec{v}

$$D_{\vec{v}}f(p) = \nabla f(p) \cdot \vec{v}$$

For vector valued functions, we can compute using the Jacobian $D_{\vec{v}}f(p) = DH(p) \cdot \vec{v}$

Definition: The directional derivative of f at $\vec{a} = (a, b)$ in the direction of \vec{v} is given by $D_{\vec{v}}f(\vec{a}) = \lim_{h \rightarrow 0} \frac{f(\vec{a} + h\vec{v}) - f(\vec{a})}{h}$, if it exists

Example: let $f(x, y) = x^2y - 3x$, $D_{\vec{v}}f(0, 0) = ?$ where $\vec{v} = \left(\frac{1}{\sqrt{2}}, \frac{-1}{\sqrt{2}}\right)$

$$D_{\vec{v}}f(0, 0) = \lim_{h \rightarrow 0} \frac{f(0, 0) + h\left(\frac{1}{\sqrt{2}}, -\frac{1}{\sqrt{2}}\right) - f(0, 0)}{h}$$

Simplify, then plug in h

$$= -\frac{3}{\sqrt{2}}$$

6.3 Multivariable differentiability at (a, b)

Definition: $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ is differentiable at (a, b) if $\exists h(x, y) = f(a, b) + f_x(a, b)x + f_y(a, b)y$

1. $f_x(a, b)$ and $f_y(a, b)$ exists

2. $\exists \mathbf{R} f'(a)$ s.t. $\lim_{h \rightarrow 0} \frac{f(x) - h(x, y)}{|x - a|} = 0$, where $h(x, y)$ is the equation of the tangent plane (or line) $f(a, b) + f_x(a, b)(x - a) + f_y(a, b)y - b$

How?

Single variable differentiability is defined by $f'(a) = \lim_{x \rightarrow a} \frac{f(x) - f(a)}{x - a}$

We can rearrange to emphasize linear approximation: $\lim_{x \rightarrow a} \frac{f(x) - [f(a) + f'(a)(x - a)]}{x - a} = 0$

This is saying that the function is differentiable at a if it can be approximated by the linear function $h(x, y)$ with error smaller than order $|x - a|$

Multivariable differentiability is now as follows $\lim_{(x, y) \rightarrow (a, b)} \frac{f(x, y) - h(x, y)}{\|(x, y) - (a, b)\|} = 0$

7 Gradients, More Derivatives, and the Jacobian

7.1 Gradient

The gradient of a scalar function is a vector that collects all the partial derivatives of f with respect to each variable:

$$\nabla f = (f_{x_1}, f_{x_2}, \dots, f_{x_n})$$

At a specific point, the gradient becomes:

$$\nabla f(\vec{a}) = (f_{x_1}(\vec{a}), \dots, f_{x_n}(\vec{a}))$$

This vector points in the direction of the steepest increase of f and its magnitude gives the rate of increase

The difference vector:

$$\vec{x} - \vec{a} = (x_1 - a_1, \dots, x_n - a_n)$$

The linear approximation of f near \vec{a} can be written as:

$$\nabla f(\vec{a})(\vec{x} - \vec{a}) = f_{x_1}(\vec{a})(x_1 - a_1) + \dots + f_{x_n}(\vec{a})(x_n - a_n)$$

Example:

Let $f(x, y) = xy^2 + e^{xy}$, find the gradient at $(0, 0)$

$$f_x = y^2 + ye^{xy}, f_y = 2yx + xe^{xy}$$

$$\nabla f = (f_x, f_y) = (y^2 + ye^{xy}, 2xy + xe^{xy}) \quad \nabla f(0, 0) = (0, 0)$$

Dot product of two vectors

If $\vec{a} = (a_1, \dots, a_n)$ and $\vec{b} = (b_1, \dots, b_n)$, then $\vec{a} \cdot \vec{b} = a_1b_1 + \dots + a_nb_n$

7.2 Derivative Matrix

Let $U \subseteq \mathbf{R}^n$ and $f : U \subseteq \mathbf{R}^n \rightarrow \mathbf{R}^m$

$$f = (f_1, f_2, \dots, f_m)$$

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

$$f_1(x) = x^2, f_2(x) = x + y$$

$$Df = \begin{matrix} \nabla f_1 \\ \nabla f_2 \\ \dots \\ \nabla f_m \end{matrix} = \begin{bmatrix} \frac{\partial f_1}{\partial x_1}, \frac{\partial f_1}{\partial x_2}, \dots, \frac{\partial f_1}{\partial x_n} \\ \frac{\partial f_2}{\partial x_1}, \frac{\partial f_2}{\partial x_2}, \dots, \frac{\partial f_2}{\partial x_n} \\ \dots \\ \frac{\partial f_m}{\partial x_1}, \frac{\partial f_m}{\partial x_2}, \dots, \frac{\partial f_m}{\partial x_n} \end{bmatrix}$$

This matrix is called the matrix of partial derivatives of f , otherwise called the Derivative Matrix or the **Jacobian Matrix**. Essentially, the derivative is a linear map, and in coordinates it is built from the partial derivatives

Example:

Let $f(x, y) = (xy, y^2 \sin x, x^3 e^y)$, find the derivative matrix

$$Df = \begin{pmatrix} \nabla f_1 & y, x \\ \nabla f_2 & y^2 \cos x, 2y \sin x \\ \nabla f_3 & 3x^2 e^y, x^3 e^y \end{pmatrix}$$

7.3 Differentiability in higher dimensions $f : U \rightarrow \mathbf{R}^m$

f is differentiable if: - $Df(\vec{a})$ exists - Tangent plane $h : \mathbf{R}^n \rightarrow \mathbf{R}^m$, $h(\vec{x}) = f(\vec{a}) + Df(\vec{a})(\vec{x} - \vec{a})$, where $Df(\vec{a})(\vec{x} - \vec{a})$ is a matrix multiplication, satisfies $\lim_{\vec{x} \rightarrow \vec{a}} \frac{\|f(\vec{x}) - h(\vec{x})\|}{\|\vec{x} - \vec{a}\|} = 0$, which is hard to use

This is why we introduce the following theorems:

7.3.1 Theorems for higher-dimension differentiability

Theorem 1:

If $f = (f_1, f_2, \dots, f_m)$, then f is differentiable at $\vec{a} \Leftrightarrow f_1, f_2, \dots, f_m$ is differentiable at \vec{a}

Theorem 2:

If $f = (f_1, f_2, \dots, f_m)$ and all partials $\frac{\partial f_i}{\partial x_j}$, as i, j, \dots, i_m, j_m , are continuous then f is differentiable

Example:

$f(x, y) = (x^2 y, e^y \sin x)$ is differentiable because all of its partial derivatives are continuous

Theorem 3:

If f is differentiable at \vec{a} , then directional derivatives can be computed using: $D_{\vec{v}}f(\vec{a}) = \nabla f(\vec{a}) \cdot \vec{v}$

If f is differentiable at \vec{a} , then $D_{\vec{v}}f(\vec{a}) = Df(\vec{a})\vec{v}$ where $Df(\vec{a})\vec{v}$ is a matrix multiplication

Example:

$f(x, y) = (e^x y, x^2 y)$, find rate of change of f at $(1, 2)$ in direction $\vec{v} = \left(-\frac{1}{2}, \frac{\sqrt{3}}{2}\right)$

$$Df = \begin{pmatrix} e^x y & e^x \\ 2xy & x^2 \end{pmatrix}, Df(1, 2) = \begin{pmatrix} 2e & e \\ 4 & 1 \end{pmatrix}$$

$$Df(1, 2)\vec{v} = \begin{pmatrix} 2e & e \\ 4 & 1 \end{pmatrix} \cdot \begin{pmatrix} -\frac{1}{2} \\ \frac{\sqrt{3}}{2} \end{pmatrix} = \begin{pmatrix} e + \frac{\sqrt{3}}{2}e \\ 2 + \frac{\sqrt{3}}{2} \end{pmatrix}$$

7.4 Properties of Differentiability

Let $F : \mathbf{R}^n \rightarrow \mathbf{R}, G : \mathbf{R}^n \rightarrow \mathbf{R}$ be differentiable at \vec{a}

- $F + G$ is differentiable at \vec{a}
- $F \cdot G$ is differentiable at \vec{a}
- If $G(\vec{a}) \neq 0$, $\frac{F}{G}$ is differentiable at \vec{a}
- If f is differentiable at a and g is differentiable at $f(a)$, then $g \circ f$ is differentiable at a and $\frac{d}{dx}(g \circ f) = g'(f(a)) * f'(a)$

- The graph of a function is the set $\{(x, y, f(x, y)) \in \mathbf{R}^3 : (x, y) \in \text{domain}\}$
- If f_x, f_y, f_{xy}, f_{yx} are continuous, then $f_{xy} = f_{yx}$

8 Differentiability in higher dimension

8.1 Chain Rule in Composition

$D(G \circ F)(\vec{a}) = DG(F(\vec{a}))DF(\vec{a})$, where the RHS is a matrix multiplication

Example: $F(x, y) = (x^2y, e^{3x})$ and $G(x, y) = (x + y, xy, \sin(2x - y))$

Find: $D(G \circ F)(1, 1)$, where $(1, 1) = (\vec{a})$

Apply the chain rule equation and get $= DG(1, e^3)DF(1, 1)$

$$DF = \begin{pmatrix} 2xy & x^2 \\ 3e^{3x} & 0 \end{pmatrix} \text{ and } DG = \begin{pmatrix} 1 & 1 \\ y & x \\ 2\cos(2x - y) & -\cos(2x - y) \end{pmatrix}$$

$$DF(1, 1) = \begin{pmatrix} 2 & 1 \\ 3e^3 & 0 \end{pmatrix} \text{ and } DG(1, e^3) = \begin{pmatrix} 1 & 1 \\ e^3 & 1 \\ 2\cos(2 - e^3) & -\cos(2 - e^3) \end{pmatrix}$$

$$\text{Now, } D(G \circ F)(1, 1) = \begin{pmatrix} 2 + 3e^3 & 1 \\ 5e^3 & e^3 \\ 4\cos(2 - e^3) - 3e^3\cos(2 - e^3) & 2\cos(2 - e^3) \end{pmatrix}$$

8.2 Polar Coordinate Examples

$$x = r \cos \theta, y = r \sin \theta$$

$$DH(r, \theta) = DG(r \cos \theta, r \sin \theta)DF(r, \theta)$$

$$DH(r, \theta) = \frac{\partial G}{\partial x} \cos \theta + \frac{\partial G}{\partial y} \sin \theta - \frac{\partial G}{\partial x} r \sin \theta + \frac{\partial G}{\partial y} \cos \theta$$

Example: Find DH

With a given $r, \theta, \frac{\partial G}{\partial x}, \frac{\partial G}{\partial y}$, we can find $DH(r, \theta)$ through the chain rule

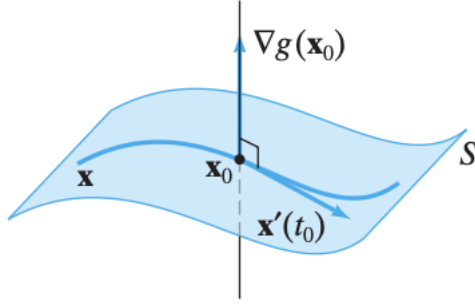
Example: Find DG

With a given $r, \theta, \frac{\partial H}{\partial x}, \frac{\partial H}{\partial \theta}$, we can find DG with: $\left[\frac{\partial G}{\partial x}, \frac{\partial G}{\partial y} \right] = \left[\frac{\partial H}{\partial x}, \frac{\partial H}{\partial \theta} \right] \cdot DF^{-1}$

9 Applications of the Gradient

9.1 Gradients and level curves

If we have a level curve for the function $x^2 + y^2$, so $f(x, y) = c = x^2 + y^2$, then the gradient ∇F is always perpendicular to the tangent plane to the level curve



Thus, the equation of the tangent plane is given by $\nabla F \cdot (\vec{x} - \vec{a}) = 0, \forall \vec{x}$ on tangent plane, where \vec{a} is the fixed reference vector

Example: Find equation of tangent plane given the function and the reference vector

$$f(x, y) = x^2y + ye^x \text{ at } (0, 1, -1)$$

Isolate and get the gradient: $f(x, y, z) = z - x^2y + ye^x$ $\nabla F = (-2xy + ye^x, -x^2 + e^x, 1)$
 $\nabla F(0, 1, -1) = (1, 1, 1)$

$$(1, 1, 1) \cdot (x - 0, y - 1, z + 1) = 0 \therefore x + y + z = 0$$

9.2 Magnitude of ∇F

Consider the directional derivative $D_{\vec{v}}f(\vec{a}) = \nabla f(\vec{a}) \cdot \vec{v}$

In what direction does the function increase the most?

If θ is the angle between \vec{v} and the gradient vector $\nabla f(\vec{a})$, then we have:

$$D_{\vec{v}}f(\vec{a}) = \|\nabla f(\vec{a})\| \|\vec{v}\| \cos \theta = \|\nabla f(\vec{a})\| \cos \theta \text{ because the magnitude of the unit vector } \vec{v} = 1$$

Thus, the max ROC is at $\theta = 0, = \|\nabla f(\vec{a})\|$

The min ROC is at $\theta = \pi, = -\|\nabla f(\vec{a})\|$ and is opposite to $\nabla f(\vec{a})$

9.2.1 Example

Given $f(x, y) = 3 \sin xy, \vec{a} = (1, \pi)$ find: 1. direction of max ROC, value of ROC at $f(\vec{a})$, and direction of tangent to the level curve at \vec{a}

1. Get gradient, plug in point, \therefore max ROC is in the direction of gradient
2. Get magnitude of gradient at point, \therefore this is the max ROC
3. ∇f is perpendicular to tangent line to the level curve at $(1, \pi)$. Find $\vec{v} \perp (-3\pi, -3)$

Method: change values in vector, change sign of 1

$\vec{v}_1 = (3, -3\pi)$ SOLVE USING CHAT

10 Conservative Vector Fields

A vector field is conservative if $\exists f : U \rightarrow \mathbf{R}$ such that $F = \nabla f$

The function f is called a potential function of F

Example: $F(x, y) = (2x, 2y)$

Thus, if $F = \nabla f$ and the potential function $f(x, y) = x^2 + y^2$, then $F(x, y)$ is conservative and f is the potential function

10.1 Test for conservative

Function $G(x, y, z)$ is conservative if

$$\begin{array}{ccc}
 (G_1)_y = (G_2)_x & (G_2)_z = (G_3)_y & (G_1)_z = (G_3)_x \\
 \parallel & \parallel & \parallel \\
 F_{xy} = F_{yx} & F_{yz} = F_{zy} & F_{xz} = F_{zx}
 \end{array}$$

10.2 Reconstruct a potential function given its gradient

Find $\nabla f = (f_x, f_y, f_z) = g = (g_1, g_2, g_3)$

1. Integrate g_1 wrt x

$$f(x, y, z) = \int g_1 dx + h(y, z)$$

2. Differentiate wrt y , set equal to g_2 , solve for $h(y, z)$ by integrating wrt y and get a $k(z)$ term
3. Differentiate wrt z , set equal to g_3 , solve for $k(z)$ up to constant C
4. Assemble final $f(x, y, z) + C$

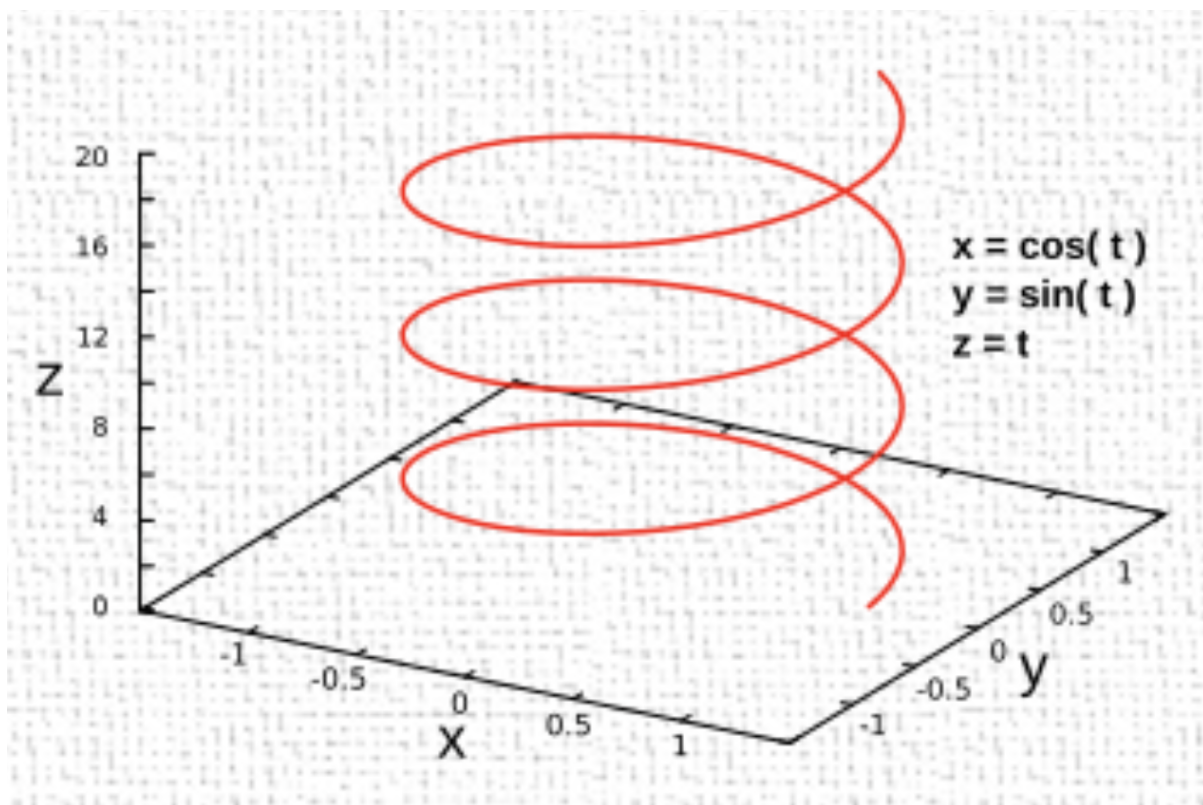
11 Parametric Equations and Class

Definition of Path: a continuous function $f : I \rightarrow \mathbf{R}^n$ where $I \in \mathbf{R}$ is on the interval $[a, b]$

11.1 Parametrization

$f(a)$ = starting point of f , $f(b)$ = end point of f

The Im of the path, denoted by $f(I)$ is called the curve in \mathbf{R}^2 and f is a parametrization of C



Important result: Parametrization is not unique

$f(t) = (\cos t, \sin t)$ and $g(t) = (t, \sqrt{1-t^2})$ have the same curve $\text{Im}(f) = \text{Im}(g)$

11.2 Class

Let $f : I \rightarrow \mathbf{R}^n$ be a path, say f is of class $C^{(k)}$, $k \in \mathbf{N}$, and f is differentiable k -times and *derivatives are continuous*

Example: $y^2 = x^3$

Parametrized: $f(t) = (t, t^{3/2}) \rightarrow f'(t) = (1, \frac{3}{2} \cdot \sqrt{t}) \rightarrow f'' = (0, \frac{3}{4} \cdot \frac{1}{\sqrt{t}})$, which is not defined at $t = 0$

$\therefore f$ is of class C^1 and not C^2

12 Arc Length, Divergence and Curl

12.1 Arc Length

Arc length from a to b , with $f : I \rightarrow \mathbb{R}^m$, and c is a curve in f :

$$L(f) = \int_a^b \|f'(t)\| dx$$

Method: get parametrization $f(t)$, get speed, then integrate w.r.t. bounds

12.2 Divergence of a vector field

Denoted by $Div(f) = \nabla \cdot \vec{F} = \frac{\partial F_x}{\partial x} + \frac{\partial F_y}{\partial y} + \frac{\partial F_z}{\partial z}$, it can measure net mass flow or flux density

If $Div(f) > 0$, consider the field as a source, flowing out If $Div(f) < 0$, consider the field as a sink, flows in

12.3 Curl of a vector field

Let $F = (F_1, F_2, F_3)$ be a differentiable vector field in \mathbb{R}^3

$$Curl(F) = \nabla \times F = \begin{vmatrix} i & -j & k \\ \frac{\partial}{\partial x} & \frac{\partial}{\partial y} & \frac{\partial}{\partial z} \\ F_1 & F_2 & F_3 \end{vmatrix}$$

$$Curl(F) = \left(\frac{\partial F_3}{\partial y} - \frac{\partial F_2}{\partial x}, -\frac{\partial F_3}{\partial x} + \frac{\partial F_1}{\partial z}, \frac{\partial F_2}{\partial x} - \frac{\partial F_1}{\partial y} \right)$$

The image shows a handwritten diagram on a grid background. On the left, there is a vertical list of labels: y , z , x , y . To the right of these labels are numbers 2, 3, 1, 2 respectively, with lines connecting them in a cycle. To the right of this is the curl formula components: $\left(\frac{\partial F_3}{\partial y} - \frac{\partial F_2}{\partial x}, \frac{\partial F_1}{\partial z} - \frac{\partial F_3}{\partial x}, \frac{\partial F_2}{\partial x} - \frac{\partial F_1}{\partial y} \right)$.

13 Gradient, Divergence and Curl (cont.)

Scalar field: $f(x, y, z)$ Vector field: $\left(\frac{\partial f}{\partial x}, \frac{\partial f}{\partial y}, \frac{\partial f}{\partial z}\right)$

∇f inputs a scalar field and outputs a vector field

$\nabla \cdot F$ inputs a vector field \vec{F} and outputs a scalar field

$\nabla \times F$ inputs a vector field \vec{F} and outputs a vector field

13.1 Identities

The curl of a gradient, $\nabla \times (\nabla f) = \vec{0}$, gradient fields are irrotational

The divergence of a curl, $\nabla \cdot (\nabla \times \vec{F}) = 0$, curl fields have no net source

The divergence of a gradient, $\nabla \cdot (\nabla f)$ is the Laplacian, Δf , a scalar field

The curl of a divergence, $\nabla \times (\nabla \cdot \vec{F})$ is undefined, divergence can't input a scalar field

The gradient of a curl, $\nabla(\nabla \times \vec{F})$ is undefined, gradient can't input a vector field

The curl of a curl, $\nabla \times (\nabla \times (F)) = \nabla(Div(F)) - \nabla^2 F$, and is defined in \mathbb{R}^3

G is conservative if $\exists f : U \rightarrow \mathbb{R}$ such that $G = \nabla F$, where F is the potential function

The dot product of two vector fields, e.g. $F \cdot G$, is a scalar field defined by $\mathbb{R}^3 \rightarrow \mathbb{R}$

If $G : U(\subseteq \mathbb{R}^2) \rightarrow \mathbb{R}^2$, so (G_1, G_2) . If $Curl(G) = 0$, then G is conservative

If $G : U(\subseteq \mathbb{R}^3) \rightarrow \mathbb{R}^3$, if G is the curl of some vector field, then $div=0$

14 Special Domains and Conservative Functions

Let $U \subseteq \mathbb{R}^3$ be an open set

U is simply connected if:

1. U is connected (any two points can be connected by a path)
2. Every loop inside U can be shrunk continuously to a point inside U

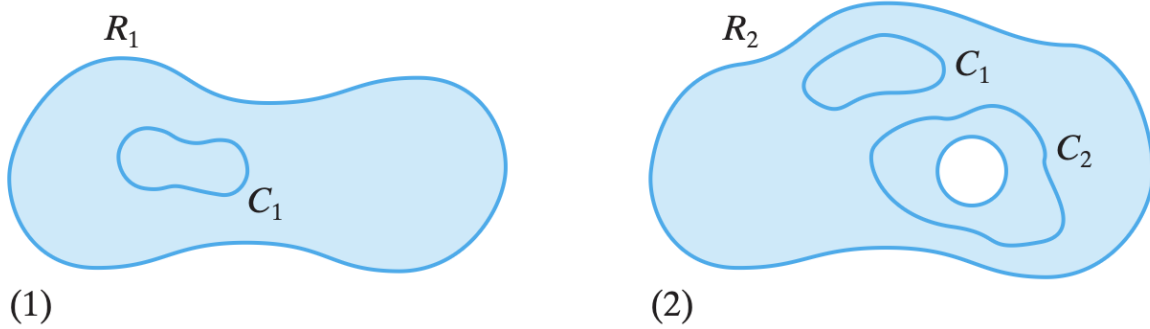


Figure 6.35 (1) The region $R_1 \subset \mathbb{R}^2$ is simply-connected: All points surrounded by any simple, closed curve in R_1 lie in R_1 . (2) In contrast, R_2 is not simply-connected: Although the curve C_1 encloses points that lie in R_2 , the curve C_2 surrounds a hole. Hence, C_2 cannot be continuously shrunk to a point while remaining in R_2 .

If we let $U \subseteq \mathbb{R}^n$ be a simply connected open set, and $F : U \rightarrow \mathbb{R}^n$ be a vector field, then f is conservative if and only if $\text{Curl}(f) = 0$

Example:

Let $G(x, y, z) = (y^2, 2xy + z, y - \sin z)$, is G conservative? If so, find the potential function f such that $G = \nabla f$

$\text{Domain}(G) = \mathbb{R}^3$, simply connected, and $\text{Curl}(G) = (1 - 1, 0, 0 - 2y - 2y) = 0$, thus G is conservative

Let $(G_1, G_2, G_3) = (F_x, F_y, F_z)$

$$F_x = y^2 \Rightarrow \int F_x dx = xy^2 + g(y, z)$$

$$F_y = 2xy + z \Rightarrow \frac{\partial F(x, y, z)}{\partial y} = 2xy + \frac{\partial g(y, z)}{\partial y} = 2xy + z \Rightarrow \frac{\partial g(y, z)}{\partial y} = z$$

$$g(y, z) = \int z dy = yz + h(z) \Rightarrow F(x, y, z) = xy^2 + yz + h(z)$$

$$F_z = y - \sin z \Rightarrow \frac{\partial F(x, y, z)}{\partial z} = y + \frac{dh(z)}{dz} = y - \sin z \Rightarrow \frac{dh(z)}{dz} = -\sin z$$

$$h(z) = \int -\sin z dz = \cos z + C$$

$$\therefore F(x, y, z) = xy^2 + yz + \cos z$$

15 Riemann Sums

15.1 Single-variable Integration

Let $f[a, b] \rightarrow \mathbb{R}$ be a function

$\int_a^b f(x) dx$ represents the area under the curve

We partition $[a, b]$ into subintervals for **Riemann sums**

Area under $f \approx$ sum of area of rectangles, $A = f(\xi_i)\Delta x_i$, $\Delta x_i = (a_i - a_{i-1})$, $\xi \in [a_{i-1}, a_i]$

A is integrable on $[a, b]$ if $\lim_{\Delta x_i \rightarrow 0} \int \sum_{i=1}^n f(\xi_i)\Delta x_i$ exists

15.2 How to integrate functions of two variables

Let $f : [a, b] \times [c, d] \rightarrow \mathbb{R}$

$$\Delta x_i = a_i - a_{i-1}, \Delta y_j = c_j - c_{j-1}$$

V of partitions = $lbh = f(\xi)\Delta x_i\Delta y_j$

$$Vol(A) \approx \sum_{i=1}^n \sum_{j=1}^m f(\xi_i)\Delta x_i\Delta y_j$$

f is integrable over $[a, b] \times [c, d]$ if $\lim_{\Delta x_i \text{ and } \Delta y_j \rightarrow 0} \sum_{i=1}^n \sum_{j=1}^m f(\xi_i)\Delta x_i\Delta y_j$ exists, and is denoted

by $\iint_{[a,b] \times [c,d]} f dA$

If f is continuous over $[a, b] \times [c, d]$, then it is integrable

Fubini's Theorem: let $f : [a, b] \times [c, d] \rightarrow \mathbb{R}$ be continuous

Then, $\iint_{[a,b] \times [c,d]} f(x, y) dx dy = \int_c^d \int_a^b f(x, y) dx dy$ and can be reversed

16 Cheat Sheet

16.1 Delta-Epsilon

The condition $0 < \sqrt{(x-a)^2 + (y-b)^2} < \delta$ means our input point is inside the δ -neighbourhood of (a, b)

The proof then shows that whenever the input point is that close to (a, b) , the function value $f(x, y)$ lies in the ε -neighbourhood of the limit L : $|f(x, y) - L| < \varepsilon$

Proof: Given $\varepsilon > 0$ We want to find $\delta > 0$ such that if $0 < \|x-a\| < \delta$, then $|f(x)-L| < \varepsilon$

Start with $|f(x) - L|$ and manipulate it to relate it to $\|x - a\|$ For instance, show: $|f(x) - L| \leq c\|x - a\|$ for some $c > 0$

Choose $\delta = \frac{\varepsilon}{c}$ and show that $|f(x) - L| < c\|x - a\| < c\delta = \varepsilon$

Therefore, $\lim_{x \rightarrow a} f(x) = L$

16.2 Disproving a Multivariable Limit

1. Prove with direct substitution

If you get a determinate value (like 5, 0, or ∞) and the function is built from continuous functions, you're done

If you get an indeterminate form like $0/0$, proceed with next steps.

2. Disprove with two-path test

For a limit approaching $(0, 0)$, common paths to test include: axis paths (along x , let $y = 0$, vice-versa), linear paths $y = mx$ and the limit d/n exist if it depends on m , parabolic paths

3. Disprove with polar coordinates $x = r \cos \theta, y = r \sin \theta$

16.3 Partial Derivative

Definition of partial derivatives at a point

$$\frac{\partial F}{\partial x}(0, 0) = \lim_{h \rightarrow 0} \frac{F(h, 0) - F(0, 0)}{h}$$

16.4 Derivative Matrix

$$D(G \circ F)(\vec{a}) = DG(F(\vec{a}))DF(\vec{a})$$

$$Df = \begin{bmatrix} \nabla f_1 \\ \nabla f_2 \\ \vdots \\ \nabla f_m \end{bmatrix} = \begin{bmatrix} \frac{\partial f_1}{\partial x_1}, \frac{\partial f_1}{\partial x_2}, \dots, \frac{\partial f_1}{\partial x_n} \\ \frac{\partial f_2}{\partial x_1}, \frac{\partial f_2}{\partial x_2}, \dots, \frac{\partial f_2}{\partial x_n} \\ \vdots \\ \frac{\partial f_m}{\partial x_1}, \frac{\partial f_m}{\partial x_2}, \dots, \frac{\partial f_m}{\partial x_n} \end{bmatrix}$$

$$\text{Let } A = [a, b] \text{ and } B = \begin{bmatrix} e & f \\ g & h \end{bmatrix}, \text{ then } AB = [ae + bg \quad af + bh]$$

Let A be of size $m \times m$ and B of size $p \times q$, then $C = A \times B$ has dimensions $m \times q$

16.5 Divergence and Curl

The divergence of F denoted by $\nabla \cdot F$ is $\mathbb{R}^3 \rightarrow \mathbb{R}$, measures the net rate of flow outward from a point, and is $\nabla \cdot F = \frac{\partial F_1}{\partial x} + \frac{\partial F_2}{\partial y} + \frac{\partial F_3}{\partial z}$

The curl of F denoted by $\nabla \times F$ is $\mathbb{R}^3 \rightarrow \mathbb{R}^3$, measures the tendency to rotate or swirl around a point, and is $\nabla \times F = \left\langle \frac{\partial F_3}{\partial y} - \frac{\partial F_2}{\partial z}, \frac{\partial F_1}{\partial z} - \frac{\partial F_3}{\partial x}, \frac{\partial F_2}{\partial x} - \frac{\partial F_1}{\partial y} \right\rangle$

The gradient of f denoted by ∇f is $\mathbb{R}^3 \rightarrow \mathbb{R}^3$ points in the direction of greatest increase of f , and its magnitude is the rate of increase, and is $\nabla f = \left\langle \frac{\partial f}{\partial x}, \frac{\partial f}{\partial y}, \frac{\partial f}{\partial z} \right\rangle$

$\nabla \cdot (\nabla \times F) = 0$, or in words, the divergence of the curl of any vector field F is 0