

MATH 20800 (Honors Analysis in \mathbb{R}^n II) Notes

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Chapter 1

The Algebra and Topology of \mathbb{R}^n

1.1 Notes

- 1/10:
- Syllabus.
 - In his mind, homework is the main setting where learning takes place.
 - We're going to be studying analysis, or calculus, on **manifolds** this quarter.
 - **Manifold**: A “space” that looks like Euclidean space \mathbb{R}^n locally.
 - The surfaces of a sphere and torus are common examples of 2-dimensional manifolds.
 - With regard to the above definition, think about how people in ancient times didn't think the Earth was a sphere because it looked like a plane locally.
 - This class will look much like a calculus course, in that we first talk about limits, then differentiation, then integration, and culminating in the fundamental theory of calculus.
 - Last quarter, we primarily developed linear algebra and basic topology on metric spaces.
 - Chapter 1 of Munkres (1991) is a review of what's needed from last quarter.
 - This is all basically continuity.
 - Thus, we can start right up with differentiation.

1.2 The Algebra and Topology of \mathbb{R}^n

From Munkres (1991).

- 1/17:
- “In the first part of this book, \mathbb{R}^n and its subspaces are the only vector spaces with which we shall be concerned. In later chapters, we shall deal with more general vector spaces” (Munkres, 1991, p. 2).
 - **Inner product**: Denoted by $\langle \mathbf{x}, \mathbf{y} \rangle$.
 - **Euclidean norm**: The following norm. Denoted by $\|\mathbf{x}\|$. Given by

$$\|\mathbf{x}\| = \sqrt{x_1^2 + \cdots + x_n^2}$$

- **Sup norm** (of n -tuples): The following norm. Denoted by $|\mathbf{x}|$. Given by

$$|\mathbf{x}| = \max\{|x_1|, \dots, |x_n|\}$$

- Note that the Euclidean norm and sup norm satisfy the inequalities

$$|\mathbf{x}| \leq \|\mathbf{x}\| \leq \sqrt{n}|\mathbf{x}|$$

for all $\mathbf{x} \in \mathbb{R}^n$.

- **Sup norm** (of matrices): The following norm. *Denoted by $|\mathbf{A}|$. Given by*

$$|\mathbf{A}| = \max\{|a_{ij}| : i \in [n], j \in [m]\}$$

- Theorem 1.3: If A has size n by m and B has size m by p , then

$$|\mathbf{A} \cdot \mathbf{B}| \leq m|\mathbf{A}| |\mathbf{B}|$$

- **Echelon form**: *Also known as **stairstep form**.*

- **Transpose** (of A): *Denoted by \mathbf{A}^{tr} .*

- Theorem 2.1: Let A be an n -by- m matrix. Any elementary row operation on A may be carried out by premultiplying A by the corresponding elementary matrix.

– “We will use this result later on when we prove the change of variables theorem for a multiple integral” (Munkres, 1991, p. 12).

- **Determinant** (of A): *Denoted by $\det \mathbf{A}$. Not denoted by $|\mathbf{A}|$.*

- **Determinant function**: A function that assigns to each n -by- n matrix A a real number denoted $\det A$ and satisfies the following axioms.

1. If B is the matrix obtained by exchanging any two rows of A , then $\det B = -\det A$.
2. Given i , the function $\det A$ is linear as a function of the i^{th} row alone.
3. $\det I_n = 1$.

- Corollary 2.9: The determinant function is uniquely characterized by its three axioms.

- **ϵ -neighborhood** (of x_0): The following set, where X is a metric space with metric d , $x_0 \in X$, and $\epsilon > 0$. *Also known as **ϵ -neighborhood centered at x_0** . Denoted by $U(\mathbf{x}_0; \epsilon)$. Given by*

$$U(x_0; \epsilon) = \{x \mid d(x, x_0) < \epsilon\}$$

- **Topological property** (of X): A property of a metric space X that depends only on the collection of open sets of X , rather than on the specific metric involved.

– Examples include limits, continuity, and compactness.

- **Interior** (of $A \subset \mathbb{R}^n$): The union of all open sets of \mathbb{R}^n that are contained in A . *Denoted by $\text{Int } \mathbf{A}$.*

- **Exterior** (of $A \subset \mathbb{R}^n$): The union of all open sets of \mathbb{R}^n that are disjoint from A . *Denoted by $\text{Ext } \mathbf{A}$.*

- **Boundary** (of $A \subset \mathbb{R}^n$): The set of all points of \mathbb{R}^n that are contained in neither $\text{Int } A$ nor $\text{Ext } A$. *Denoted by $\text{Bd } \mathbf{A}$.*

– $\mathbf{x} \in \text{Bd } A$ iff every open set containing \mathbf{x} intersects both A and $\mathbb{R}^n \setminus A$.

Chapter 2

Differentiation

2.1 Notes

1/10:

- Since manifolds look like Euclidean spaces locally, we basically only need to study differentiation on Euclidean spaces.
- Set up: Let $U \subset \mathbb{R}^n$ be open, and $f : U \rightarrow \mathbb{R}^m$ be a function.
- Idea: The derivative of f at some point $\mathbf{a} \in U$ is “the best linear approximation” to f at \mathbf{a} .
- **Differentiable** (function f at \mathbf{a}): A function f for which there exists a linear transformation $A : \mathbb{R}^n \rightarrow \mathbb{R}^m$ such that

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

- **Total derivative** (of f at \mathbf{a}): The linear transformation A corresponding to a differentiable function f . Denoted by $Df(\mathbf{a})$.
- Questions to ask:
 1. When does the total derivative exist?
 2. When it does exist, can there be multiple?
 3. When it exists and is unique, how do I calculate it?
- Proposition: If A, B are linear transformations that both satisfy the definition, then $A = B$.
 - We have

$$\lim_{\mathbf{h} \rightarrow \mathbf{0}} \frac{f(\mathbf{a} + \mathbf{h}) - f(\mathbf{a}) - A\mathbf{h}}{\|\mathbf{h}\|} = \mathbf{0} \qquad \lim_{\mathbf{h} \rightarrow \mathbf{0}} \frac{f(\mathbf{a} + \mathbf{h}) - f(\mathbf{a}) - B\mathbf{h}}{\|\mathbf{h}\|} = \mathbf{0}$$

- It follows by subtracting the right equation above from the left one that

$$\lim_{\mathbf{h} \rightarrow \mathbf{0}} \frac{A\mathbf{h} - B\mathbf{h}}{\|\mathbf{h}\|} = \mathbf{0}$$

- Apply linearity: For an arbitrary $\mathbf{v} \in \mathbb{R}^n$ and $t \in \mathbb{R}, t > 0$, we have

$$\frac{A(t\mathbf{v}) - B(t\mathbf{v})}{t} = A\mathbf{v} - B\mathbf{v}$$

- Therefore, since $t\mathbf{v} \rightarrow \mathbf{0}$ as $t \rightarrow 0$, we have by the above that

$$\begin{aligned}\mathbf{0} &= \lim_{t \rightarrow 0} \frac{A(t\mathbf{v}) - B(t\mathbf{v})}{\|t\mathbf{v}\|} \\ &= \lim_{t \rightarrow 0} \frac{A\mathbf{v} - B\mathbf{v}}{\|\mathbf{v}\|} \\ \mathbf{0} \cdot \|\mathbf{v}\| &= \lim_{t \rightarrow 0} (A\mathbf{v} - B\mathbf{v}) \\ \mathbf{0} &= A\mathbf{v} - B\mathbf{v} \\ B\mathbf{v} &= A\mathbf{v}\end{aligned}$$

- Example: Let $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$ be linear, i.e., $f(\mathbf{v}) = A\mathbf{v}$ for some linear transformation A . Then for all $\mathbf{a} \in \mathbb{R}^n$, $Df(\mathbf{a}) = A$ is constant.

- We have from the definition that

$$\begin{aligned}\lim_{\mathbf{h} \rightarrow \mathbf{0}} \frac{f(\mathbf{a} + \mathbf{h}) - f(\mathbf{a}) - A\mathbf{h}}{\|\mathbf{h}\|} &= \lim_{\mathbf{h} \rightarrow \mathbf{0}} \frac{f(\mathbf{a}) + f(\mathbf{h}) - f(\mathbf{a}) - f(\mathbf{h})}{\|\mathbf{h}\|} \\ &= \lim_{\mathbf{h} \rightarrow \mathbf{0}} \frac{\mathbf{0}}{\|\mathbf{h}\|} \\ &= \mathbf{0}\end{aligned}$$

- Theorem: If f is differentiable at \mathbf{a} , then f is continuous at \mathbf{a} .

- By definition, there exists a linear transformation A such that

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

- Additionally, we have that

$$f(\mathbf{a} + \mathbf{h}) = f(\mathbf{a}) + A\mathbf{h} + \|\mathbf{h}\| \left(\frac{f(\mathbf{a} + \mathbf{h}) - f(\mathbf{a}) - A\mathbf{h}}{\|\mathbf{h}\|} \right)$$

- As $\mathbf{h} \rightarrow \mathbf{0}$, the right-hand side of the above equation goes to $f(\mathbf{a})$.

- As a linear transformation, $A\mathbf{h} \rightarrow \mathbf{0}$ as $\mathbf{h} \rightarrow \mathbf{0}$.
- Clearly $\|\mathbf{h}\| \rightarrow \mathbf{0}$ as $\mathbf{h} \rightarrow \mathbf{0}$.
- And we have by definition that the last term goes to $\mathbf{0}$ as $\mathbf{h} \rightarrow \mathbf{0}$.

- Therefore, f is continuous at \mathbf{a} .

- Observation: A function $f : U \rightarrow \mathbb{R}^m$ is given by an m -tuple of functions $f_1 : U \rightarrow \mathbb{R}$ known as components. $f = (f_1, \dots, f_m)$.

- Proposition: f is differentiable at $\mathbf{a} \in U$ iff each component function f_i is differentiable at \mathbf{a} . In this case,

$$Df(\mathbf{a}) = (Df_1(\mathbf{a}), \dots, Df_m(\mathbf{a}))$$

- We know that

$$\lim_{\mathbf{h} \rightarrow \mathbf{0}} \frac{f(\mathbf{a} + \mathbf{h}) - f(\mathbf{a}) - A\mathbf{h}}{\|\mathbf{h}\|} \in \mathbb{R}^m$$

- Thus, the limit is zero iff the limit of each component is zero.

- We have that the i^{th} component of the vector on the left below is equal to the number on the right; we call the common value $L_i(\mathbf{h})$.

$$\left(\frac{f(\mathbf{a} + \mathbf{h}) - f(\mathbf{a}) - A\mathbf{h}}{\|\mathbf{h}\|} \right)_i = \frac{f_i(\mathbf{a} + \mathbf{h}) - f_i(\mathbf{a}) - (A\mathbf{h})_i}{\|\mathbf{h}\|}$$

- The upshot is that f is differentiable at \mathbf{a} iff $\lim_{\mathbf{h} \rightarrow \mathbf{0}} L_i(\mathbf{h}) = \mathbf{0}$ iff the linear transformation $\mathbf{h} \mapsto (A\mathbf{h})_i : \mathbb{R}^m \rightarrow \mathbb{R}$ is the total derivative of f_i .

- Now, each f_i is a function of n variables, i.e., $f_i(x_1, \dots, x_n)$ where x_1, \dots, x_n are coordinates on \mathbb{R}^n .

1/12:

- **Partial derivative** (of f wrt. x_i at $\mathbf{a} \in U$): The following quantity. Denoted by $\partial f / \partial x_i$. Given by

$$\frac{\partial f}{\partial x_i} = \lim_{h \rightarrow 0} \frac{f(a_1, \dots, a_{i-1}, a_i + h, a_{i+1}, \dots, a_n) - f(\mathbf{a})}{h}$$

- The partial derivative is easy to calculate if you're good at calculating single-variable derivatives.

- Questions:

1. If the partial derivatives all exist, does the total derivative also exist?
2. If partial derivatives exist, is f continuous?

- The answer is no to both — it's too weak a condition.

- Counter example: Consider $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ given by

$$f(x, y) = \begin{cases} \frac{x^2 y}{x^4 + y^4} & (x, y) \neq \mathbf{0} \\ 0 & (x, y) = \mathbf{0} \end{cases}$$

- All partial derivatives exist at $(0, 0)$ but f is not continuous at $(0, 0)$.

- We'll consider this in the homework.

- Now we try taking derivatives in infinitely many directions, as opposed to just n many.

- **Directional derivative** (of f at \mathbf{a} in the direction of $\mathbf{v} \in \mathbb{R}^n$): The following quantity. Denoted by $D_{\mathbf{v}}f(\mathbf{a})$, $\partial f / \partial \mathbf{v}$. Given by

$$D_{\mathbf{v}}f(\mathbf{a}) = \frac{\partial f}{\partial \mathbf{v}} = \lim_{h \rightarrow 0} \frac{f(\mathbf{a} + h\mathbf{v}) - f(\mathbf{a})}{h}$$

- We always take $\|\mathbf{v}\| = 1$.

- The partial derivative is just a directional derivative along the standard basis vectors. Alternatively, the directional derivative is just a generalization of the partial derivatives.

- This still isn't a strong enough condition — the above counterexample has all directional derivatives at $(0, 0)$ but still isn't continuous.

- Proposition: Suppose f is differentiable at $\mathbf{a} \in U$. Then all directional derivatives of f at \mathbf{a} exist and for all $\mathbf{v} \in \mathbb{R}^n$,

$$\frac{\partial f}{\partial \mathbf{v}} = Df(\mathbf{a})(\mathbf{v})$$

- The total derivative says that the derivative exists from all sequences of approach. We're just going to pick a particular vector direction of approach.

- Mathematically, by the definition of the total derivative,

$$\begin{aligned} \mathbf{0} &= \lim_{h \rightarrow 0} \frac{f(\mathbf{a} + h\mathbf{v}) - f(\mathbf{a}) - Df(\mathbf{a})(h\mathbf{v})}{h} \\ &= \lim_{h \rightarrow 0} \frac{f(\mathbf{a} + h\mathbf{v}) - f(\mathbf{a})}{h} - Df(\mathbf{a})(\mathbf{v}) \\ Df(\mathbf{a})(\mathbf{v}) &= \frac{\partial f}{\partial \mathbf{v}} \end{aligned}$$

- A particular consequence is that

$$\frac{\partial f}{\partial x_i} = Df(\mathbf{a})(e_i)$$

- But the total derivative, as a linear transformation, is completely defined by its behavior on the basis vectors.
- Thus, it is defined by the m -by- n matrix

$$Df(\mathbf{a}) = \left(\frac{\partial f_j}{\partial x_i} \right)_{\substack{1 \leq j \leq m \\ 1 \leq i \leq n}}$$

- **Jacobian matrix** (of f at \mathbf{a}): The above matrix, representing the total derivative of f at \mathbf{a} .
- Theorem: Suppose $f : U \rightarrow \mathbb{R}^m$ is a function on an open set $U \subset \mathbb{R}^n$. If all partial derivatives of f exist and are continuous on U , then f is differentiable on U .

- Recall the mean value theorem (MVT): Suppose $g : [a, b] \rightarrow \mathbb{R}$ is a continuous function which is differentiable on (a, b) . Then there exists $c \in (a, b)$ such that $g'(c) = [g(b) - g(a)]/[b - a]$.
- WLOG let $m = 1$ (if we prove this case, we can use the proposition relating f to its components to prove the general case).
- Rewrite

$$\begin{aligned} f(\mathbf{a} + \mathbf{h}) - f(\mathbf{a}) &= f(a_1 + h_1, a_2 + h_2, a_3 + h_3, \dots, a_n + h_n) - f(a_1, a_2 + h_2, a_3 + h_3, \dots, a_n + h_n) \\ &\quad + f(a_1, a_2 + h_2, a_3 + h_3, \dots, a_n + h_n) - f(a_1, a_2, a_3 + h_3, \dots, a_n + h_n) \\ &\quad + \dots \\ &\quad + f(a_1, \dots, a_{n-1}, a_n + h_n) - f(\mathbf{a}) \end{aligned}$$

where $\mathbf{a} = (a_1, \dots, a_n)$ and $\mathbf{h} = (h_1, \dots, h_n)$.

- Apply the MVT to each term to get

$$f(a_1, \dots, a_i + h_i, \dots, a_n + h_n) - f(a_1, \dots, a_i, \dots, a_n + h_n) = h_i \frac{\partial f}{\partial x_i}(a_1, \dots, c_i(\mathbf{h}), \dots, a_n + h_n)$$

for some $c_i(\mathbf{h}) \in (a_i, a_i + h_i) \cup (a_i + h_i, a_i)$.

- Now let A be the Jacobian matrix of f at \mathbf{a} .
- WTS:

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

- We have

$$f(\mathbf{a} + \mathbf{h}) - f(\mathbf{a}) = \sum_{i=1}^n h_i \frac{\partial f}{\partial x_i}(a_1, \dots, c_i(\mathbf{h}), \dots, a_n + h_n)$$

- Let $\pi_i : \mathbb{R}^n \rightarrow \mathbb{R}^n$ be the linear map $(x_1, \dots, x_n) \mapsto (0, \dots, x_i, \dots, 0)$. Clearly, $\mathbf{x} = \sum_{i=1}^n \pi_i \mathbf{x}$.
- Thus, $A\mathbf{h} = \sum_{i=1}^n A\pi_i \mathbf{h}$ and $A\pi_i \mathbf{h} = \frac{\partial f}{\partial x_i}(\mathbf{a}) \cdot h_i$.
- Applying, we have

$$\begin{aligned} \frac{f(\mathbf{a} + \mathbf{h}) - f(\mathbf{a}) - A\mathbf{h}}{\|\mathbf{h}\|} &= \sum_{i=1}^n \frac{1}{\|\mathbf{h}\|} \left(h_i \frac{\partial f}{\partial x_i}(a_1, \dots, a_{i-1}, c_i(\mathbf{h}), a_{i+1} + h_{i+1}, \dots, a_n + h_n) - \frac{\partial f}{\partial x_i}(\mathbf{a}) \cdot h_i \right) \\ &= \sum_{i=1}^n \frac{h_i}{\|\mathbf{h}\|} \left(\frac{\partial f}{\partial x_i}(a_1, \dots, a_{i-1}, c_i(\mathbf{h}), a_{i+1} + h_{i+1}, \dots, a_n + h_n) - \frac{\partial f}{\partial x_i}(\mathbf{a}) \right) \end{aligned}$$

- We know that $-1 \leq h_i/\|\mathbf{h}\| \leq 1$, so we need only show that the difference above goes to zero as $\mathbf{h} \rightarrow \mathbf{0}$. But we know this by the continuity of the partial derivatives.

- Note that this theorem gives a sufficient condition but not a necessary condition for f to be differentiable.

1/14:

- Theorem (Chain Rule): Suppose $f : U \rightarrow \mathbb{R}^m$ and $g : V \rightarrow \mathbb{R}^p$ are functions defined on open sets $U \subset \mathbb{R}^n$, $V \subset \mathbb{R}^m$ with $f(U) \subset V$. Suppose that f is differentiable at $\mathbf{a} \in U$ and g is differentiable at $\mathbf{b} = f(\mathbf{a}) \in V$. Then the composite function $g \circ f : U \rightarrow \mathbb{R}^p$ is differentiable at \mathbf{a} and $D(g \circ f)(\mathbf{a}) = Dg(\mathbf{b}) \circ Df(\mathbf{a}) : \mathbb{R}^n \rightarrow \mathbb{R}^p$.

- Note that $f : U \rightarrow \mathbb{R}^m$ is differentiable at $\mathbf{a} \in U$ with derivative A iff $f(\mathbf{a} + \mathbf{h}) = f(\mathbf{a}) + A\mathbf{h} + \tilde{f}(\mathbf{h})$ such that

$$\lim_{\mathbf{h} \rightarrow \mathbf{0}} \frac{\tilde{f}(\mathbf{h})}{\|\mathbf{h}\|} = \mathbf{0}$$

where \tilde{f} is an error function.

- We're just rearranging terms here.
- If you like, \tilde{f} is the numerator from the definition of the total derivative.
- Let $A = Df(\mathbf{a})$, $B = Dg(\mathbf{b})$. Then

$$f(\mathbf{a} + \mathbf{h}) = f(\mathbf{a}) + A\mathbf{h} + \tilde{f}(\mathbf{h})$$

so

$$\begin{aligned} (g \circ f)(\mathbf{a} + \mathbf{h}) &= g(f(\mathbf{a} + \mathbf{h})) \\ &= g(f(\mathbf{a}) + A\mathbf{h} + \tilde{f}(\mathbf{h})) \\ &= g(f(\mathbf{a})) + B(A\mathbf{h} + \tilde{f}(\mathbf{h})) + \tilde{g}(A\mathbf{h} + \tilde{f}(\mathbf{h})) \\ &= g(f(\mathbf{a})) + BA\mathbf{h} + B\tilde{f}(\mathbf{h}) + \tilde{g}(A\mathbf{h} + \tilde{f}(\mathbf{h})) \end{aligned}$$

- WTS: $\lim_{\mathbf{h} \rightarrow \mathbf{0}} [B\tilde{f}(\mathbf{h}) + \tilde{g}(A\mathbf{h} + \tilde{f}(\mathbf{h}))]/\|\mathbf{h}\| = \mathbf{0}$.
- For the first half of the fraction,

$$\frac{B\tilde{f}(\mathbf{h})}{\|\mathbf{h}\|} = B \left(\frac{\tilde{f}(\mathbf{h})}{\|\mathbf{h}\|} \right) \rightarrow \mathbf{0}$$

as $\mathbf{h} \rightarrow \mathbf{0}$ since the argument goes to $\mathbf{0}$ as $\mathbf{h} \rightarrow \mathbf{0}$ and B is a linear transformation (in particular, $B(\mathbf{0}) = \mathbf{0}$).

- For the second half of the fraction,

$$\lim_{\mathbf{h} \rightarrow \mathbf{0}} \frac{\tilde{g}(A\mathbf{h} + \tilde{f}(\mathbf{h}))}{\|\mathbf{h}\|} = \lim_{\mathbf{h} \rightarrow \mathbf{0}} \frac{\tilde{g}(A\mathbf{h} + \tilde{f}(\mathbf{h}))}{\|A\mathbf{h} + \tilde{f}(\mathbf{h})\|} \cdot \frac{\|A\mathbf{h} + \tilde{f}(\mathbf{h})\|}{\|\mathbf{h}\|}$$

- The left fraction on the right side of the equality goes to zero as $\mathbf{h} \rightarrow \mathbf{0}$ by the definition of \tilde{g} .
- The right fraction on the right side of the equality is bounded since

$$\frac{\|A\mathbf{h} + \tilde{f}(\mathbf{h})\|}{\|\mathbf{h}\|} \leq \frac{\|A\mathbf{h}\|}{\|\mathbf{h}\|} + \frac{\|\tilde{f}(\mathbf{h})\|}{\|\mathbf{h}\|} \leq \|A\| + \frac{\|\tilde{f}(\mathbf{h})\|}{\|\mathbf{h}\|}$$

where $\|A\|$ is the operator norm and $\|\tilde{f}(\mathbf{h})\|/\|\mathbf{h}\| \rightarrow 0$ as $\mathbf{h} \rightarrow \mathbf{0}$ by the definition of \tilde{f} .

- Thus, the second half of the fraction goes to zero as well.
- Theorem: Let $U \subset \mathbb{R}^m$ be an open subset.

1. Suppose $f, g : U \rightarrow \mathbb{R}^m$ are functions that are differentiable at $\mathbf{a} \in U$. Then $f + g$ is also differentiable at $\mathbf{a} \in U$ and

$$D(f + g)(\mathbf{a}) = Df(\mathbf{a}) + Dg(\mathbf{a})$$

2. Suppose $f, g : U \rightarrow \mathbb{R}$ are both differentiable at $\mathbf{a} \in U$. Then $f \cdot g$ is also differentiable at \mathbf{a} , and

$$D(f \cdot g)(\mathbf{a}) = Df(\mathbf{a}) \cdot g(\mathbf{a}) + f(\mathbf{a}) \cdot Dg(\mathbf{a})$$

3. Suppose $f : U \rightarrow \mathbb{R}$ is differentiable at $\mathbf{a} \in U$ and $f(\mathbf{a}) \neq 0$. Then $1/f$ is differentiable at $\mathbf{a} \in U$ and

$$D(1/f)(\mathbf{a}) = -\frac{Df(\mathbf{a})}{f(\mathbf{a})^2}$$

- Proof of 1: Consider the functions $F : U \rightarrow \mathbb{R}^{2m}$ and $G : \mathbb{R}^m \times \mathbb{R}^m \rightarrow \mathbb{R}^m$ defined by

$$F(\mathbf{x}) = (f(\mathbf{x}), g(\mathbf{x})) \qquad G(\mathbf{y}, \mathbf{z}) = \mathbf{y} + \mathbf{z}$$

so that

$$f + g = G \circ F$$

- F is differentiable because its components are differentiable.
- G is differentiable because it's linear. This also implies that $DG(\mathbf{x}) = G$.
- Apply the chain rule to learn that $G \circ F$ is differentiable with derivative

$$\begin{aligned} D(f + g)(\mathbf{a}) &= D(G \circ F)(\mathbf{a}) \\ &= DG(F(\mathbf{a})) \circ DF(\mathbf{a}) \\ &= G(DF(\mathbf{a})) \\ &= G(Df(\mathbf{a}), Dg(\mathbf{a})) \\ &= Df(\mathbf{a}) + Dg(\mathbf{a}) \end{aligned}$$

- Prove the others the same way.

- Theorem (Mean Value Theorem): Suppose $f : U \rightarrow \mathbb{R}$ is differentiable for all $\mathbf{a} \in U$ and that U contains the line segment joining $\mathbf{a}, \mathbf{a} + \mathbf{h} \in U$. Then there exists $t_0 \in (0, 1)$ such that

$$f(\mathbf{a} + \mathbf{h}) - f(\mathbf{a}) = Df(\mathbf{a} + t_0\mathbf{h})(\mathbf{h})$$

- Define $\phi(t) = f(\mathbf{a} + t\mathbf{h})$ for $t \in [0, 1]$.
- Apply the usual MVT to ϕ to learn that there exists $t_0 \in (0, 1)$ such that $\phi(1) - \phi(0) = \phi'(t_0)$.
- Then using the chain rule, $\phi'(t_0) = Df(\mathbf{a} + t_0\mathbf{h})(\mathbf{h})$.

- We now discuss higher order derivatives.
- **Differentiable** (f on U): A function f that is differentiable at every $\mathbf{a} \in U$.
- If f is differentiable on U , then the total derivative gives a map $Df : U \rightarrow \mathcal{L}(\mathbb{R}^n, \mathbb{R}^m)$.
 - Note that $\mathcal{L}(\mathbb{R}^n, \mathbb{R}^m)$ is isomorphic to the set of all m -by- n matrices, and \mathbb{R}^{mn} .
- We can ask for Df to itself be differentiable. We define

$$D^2f = D(Df)$$

if it exists and, more generally,

$$D^k f = D(D^{k-1} f)$$

- **Class C^k** (function): A function $f : U \rightarrow \mathbb{R}^m$ for which $Df, \dots, D^k f$ all exist and are continuous on U .

- Note that we technically need only require that $D^k f$ exist, as this implies the existence of $Df, \dots, D^{k-1}f$.
- A function $f : U \rightarrow \mathbb{R}^m$ is of class C^k iff all partial derivatives $\partial f / \partial x_i : U \rightarrow \mathbb{R}^m$ exist and are of class C^{k-1} (this follows from the theorem relating partial derivatives and differentiability).
- **Smooth** (function): A function of class C^∞ .

2.2 Chapter 2: Differentiation

From Munkres (1991).

- 1/18: • **Directional derivative** (of f at \mathbf{a} with respect to \mathbf{u}): The following limit, where $A \subset \mathbb{R}^m$ contains a neighborhood of \mathbf{a} , $f : A \rightarrow \mathbb{R}^n$, and $\mathbf{u} \in \mathbb{R}^m$ is nonzero. Denoted by $\mathbf{f}'(\mathbf{a}; \mathbf{u})$. Given by

$$\mathbf{f}'(\mathbf{a}; \mathbf{u}) = \lim_{t \rightarrow 0} \frac{f(\mathbf{a} + t\mathbf{u}) - f(\mathbf{a})}{t}$$

- Note that it is not necessary for \mathbf{u} to be a unit vector.
- If we choose as our definition of differentiability “ f is differentiable at \mathbf{a} if $\mathbf{f}'(\mathbf{a}; \mathbf{u})$ exists for every $\mathbf{u} \neq \mathbf{0}$,” we would not have results such as differentiability implies continuity and the chain rule.
 - Thus, we need a stronger definition.
- As an alternate definition of differentiability in the one-variable case, consider the following.
- **Differentiable** (single-variable real function at a): A function $\phi : A \rightarrow \mathbb{R}$, where $A \subset \mathbb{R}$ contains a neighborhood of a , for which there exists a number λ such that

$$\frac{\phi(a+t) - \phi(a) - \lambda t}{t} \rightarrow 0 \quad \text{as } t \rightarrow 0$$

- **Derivative** (of a single-variable real function at a): The unique number λ in the above definition. Denoted by $\phi'(a)$.
- “This formulation of the definition makes explicit the fact that if ϕ is differentiable, then the linear function λt is a good approximation to the **increment function** $\phi(a+t) - \phi(a)$; we often call λt the **first-order approximation** or **linear approximation** to the increment function” (Munkres, 1991, p. 43).
- **Increment function**: The function $\phi(a+t) - \phi(a)$.
- **First-order approximation**: The function λt . Also known as **linear approximation**.
- To generalize the idea of a first-order/linear approximation to the increment function $f(\mathbf{a} + \mathbf{h}) - f(\mathbf{a})$, we take a function that is linear in the sense of linear algebra.
- Note that either the sup norm or the Euclidean norm can be used in the definition of the total derivative.
- Theorem 5.1: Let $A \subset \mathbb{R}^m$, and let $f : A \rightarrow \mathbb{R}^n$. If f is differentiable at \mathbf{a} , then all the directional derivatives of f at \mathbf{a} exist, and

$$\mathbf{f}'(\mathbf{a}; \mathbf{u}) = Df(\mathbf{a}) \cdot \mathbf{u}$$
- Theorem 5.2: Let $A \subset \mathbb{R}^m$, and let $f : A \rightarrow \mathbb{R}^n$. If f is differentiable at \mathbf{a} , then f is continuous at \mathbf{a} .
- **j^{th} partial derivative** (of f at \mathbf{a}): The directional derivative of f at \mathbf{a} with respect to the vector \mathbf{e}_j , provided this derivative exists. Denoted by $D_j f(\mathbf{a})$.

- Theorem 5.3. Let $A \subset \mathbb{R}^m$, and let $f : A \rightarrow \mathbb{R}$. If f is differentiable at \mathbf{a} , then

$$Df(\mathbf{a}) = [D_1f(\mathbf{a}) \quad \cdots \quad D_nf(\mathbf{a})]$$

- Theorem 5.4: Let $A \subset \mathbb{R}^m$, and let $f : A \rightarrow \mathbb{R}^n$. Suppose A contains a neighborhood of \mathbf{a} . Let $f_i : A \rightarrow \mathbb{R}$ be the i^{th} component function of f so that

$$f(\mathbf{x}) = \begin{bmatrix} f_1(\mathbf{x}) \\ \vdots \\ f_n(\mathbf{x}) \end{bmatrix}$$

- The function f is differentiable at \mathbf{a} if and only if each component function f_i is differentiable at \mathbf{a} .
- If f is differentiable at \mathbf{a} , then its derivative is the n -by- m matrix whose i^{th} row is the derivative of the function f_i , i.e.,

$$Df(\mathbf{a}) = \begin{bmatrix} Df_1(\mathbf{a}) \\ \vdots \\ Df_n(\mathbf{a}) \end{bmatrix}$$

or, in other words, $Df(\mathbf{a})$ is the matrix whose entry in row i and column j is $D_jf_i(\mathbf{a})$.

- “It is possible for the partial derivatives, and hence the Jacobian matrix, to exist *without* it following that f is differentiable at \mathbf{a} ” (Munkres, 1991, p. 47).
 - As per the example outlined in class.
- Special cases ($m = 1$ or $n = 1$).
 - If $f : \mathbb{R}^1 \rightarrow \mathbb{R}^3$, f is often interpreted as a parameterized curve and

$$Df(t) = \begin{bmatrix} f'_1(t) \\ f'_2(t) \\ f'_3(t) \end{bmatrix}$$

is the velocity vector of the curve.

- If $g : \mathbb{R}^3 \rightarrow \mathbb{R}^1$, g is often interpreted as a scalar field, and the vector field

$$Dg(\mathbf{x}) = [D_1g(\mathbf{x}) \quad D_2g(\mathbf{x}) \quad D_3g(\mathbf{x})]$$

is called the gradient of g .

- In this case, the directional derivative of g with respect to \mathbf{u} is written in calculus as the dot product of the vectors $\vec{\nabla} g$ and \mathbf{u} .

- Theorem 6.1 (Mean Value Theorem): If $\phi : [a, b] \rightarrow \mathbb{R}$ is continuous at each point of the closed interval $[a, b]$, and differentiable at each point of the open interval (a, b) , then there exists a point c of (a, b) such that

$$\phi(b) - \phi(a) = \phi'(c)(b - a)$$

- Theorem 6.2: Let A be open in \mathbb{R}^m . Suppose that the partial derivatives $D_jf_i(\mathbf{x})$ of the component functions of f exist at each point \mathbf{x} of A and are continuous on A . Then f is differentiable at each point of A .
- **Continuously differentiable** (function on A): A function f for which the partial derivatives $D_jf_i(\mathbf{x})$ of the component functions of f exist at each point $\mathbf{x} \in A$ and are continuous on A , where $A \subset \mathbb{R}^m$ is open. Also known as **class C^1** (function on A).

- There are differentiable functions that are not of class C^1 , but we will not concern ourselves with them.
- Theorem 6.3^[1]: Let A be open in \mathbb{R}^m , and let $f : A \rightarrow \mathbb{R}$ be a function of class C^2 on A . Then for each $\mathbf{a} \in A$,

$$D_k D_j f(\mathbf{a}) = D_j D_k f(\mathbf{a})$$

- Theorem 7.1 (Chain Rule): Let $A \subset \mathbb{R}^m$, $B \subset \mathbb{R}^n$, $f : A \rightarrow \mathbb{R}^n$, $g : B \rightarrow \mathbb{R}^p$, $f(A) \subset B$, and $\mathbf{b} = f(\mathbf{a})$. If f is differentiable at \mathbf{a} and g is differentiable at \mathbf{b} , then the composite function $g \circ f$ is differentiable at \mathbf{a} . Furthermore,

$$D(g \circ f)(\mathbf{a}) = Dg(\mathbf{b}) \cdot Df(\mathbf{a})$$

where the indicated product is matrix multiplication.

- Corollary 7.2: Let A be open in \mathbb{R}^m , and let B be open in \mathbb{R}^n . Let $f : A \rightarrow \mathbb{R}^n$ and $g : B \rightarrow \mathbb{R}^p$ with $f(A) \subset B$. If f and g are of class C^r , so is the composite function $g \circ f$.
- Theorem 7.3 (Mean Value Theorem): Let A be open in \mathbb{R}^m , and let $f : A \rightarrow \mathbb{R}$ be differentiable on A . If A contains the line segment with end points \mathbf{a} and $\mathbf{a} + \mathbf{h}$, then there is a point $\mathbf{c} = \mathbf{a} + t_0 \mathbf{h}$ with $0 < t_0 < 1$ of this line segment such that

$$f(\mathbf{a} + \mathbf{h}) - f(\mathbf{a}) = Df(\mathbf{c}) \cdot \mathbf{h}$$

- Theorem 7.4: Let A be open in \mathbb{R}^n , $f : A \rightarrow \mathbb{R}^n$, and $\mathbf{b} = f(\mathbf{a})$. Suppose that g maps a neighborhood of \mathbf{b} into \mathbb{R}^n , that $g(\mathbf{b}) = \mathbf{a}$, and $g(f(\mathbf{x})) = \mathbf{x}$ for all \mathbf{x} in a neighborhood of \mathbf{a} . If f is differentiable at \mathbf{a} and if g is differentiable at \mathbf{b} , then

$$Dg(\mathbf{b}) = [Df(\mathbf{a})]^{-1}$$

Proof. Let $i : \mathbb{R}^n \rightarrow \mathbb{R}^n$ be the identity function. It has total derivative I_n . But since $g(f(\mathbf{x})) = i(\mathbf{x})$ for all \mathbf{x} in a neighborhood of \mathbf{a} , the Chain Rule implies that

$$\begin{aligned} Dg(\mathbf{b}) \cdot Df(\mathbf{a}) &= I_n \\ Dg(\mathbf{b}) &= [Df(\mathbf{a})]^{-1} \end{aligned}$$

as desired. □

- It follows from Theorem 7.4 that for f^{-1} to be differentiable at \mathbf{a} , it is *necessary* that $Df(\mathbf{a})$ is invertible.
 - We will later prove that this condition is also *sufficient* for a function f of class C^1 to have a differentiable inverse.
- **Functional notation:** Notation such as ϕ' for a derivative.
- **Operator notation:** Notation such as $D\phi$ for a derivative.
- Munkres (1991) argues that Leibniz notation is a relic of a “time when the focus of every physical and mathematical problem was on the *variables* involved, and when *functions* as such were hardly even thought about” (p. 60).

¹See Theorem 15.3 in Labalme (2021).

References

- Labalme, S. (2021). *Calculus and Analytic Geometry (Thomas) notes*. Retrieved January 18, 2022, from <https://github.com/shadypuck/CAAGThomasNotes/blob/master/main.pdf>
- Munkres, J. R. (1991). *Analysis on manifolds*. Addison-Wesley.