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

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- 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).

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- "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

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

• 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

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

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

• Sup norm (of matrices): The following norm. Denoted by |A|. Given by

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

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

$$|A \cdot B| \le m|A||B|$$

- Echelon form: Also known as stairstep form.
- Transpose (of A): Denoted by 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 $\det A$. Not denoted by |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 $U(x_0; \epsilon)$. 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 Int 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 Ext A.
- Boundary (of $A \subset \mathbb{R}^n$): The set of all points of \mathbb{R}^n that are contained in neither Int A nor Ext A. Denoted by $\operatorname{Bd} A$.
 - $-\mathbf{x} \in \operatorname{Bd} A$ 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 $f: U \to \mathbb{R}^n$ 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: $\mathbb{R}^n \to \mathbb{R}^m$ such that

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

• Total derivative (of f at a): The linear transformation A corresponding to a differentiable function f. Denoted by Df(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

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

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

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

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

$$\mathbf{0} = \lim_{t \to 0} \frac{A(t\mathbf{v}) - B(t\mathbf{v})}{\|t\mathbf{v}\|}$$

$$= \lim_{t \to 0} \frac{A\mathbf{v} - B\mathbf{v}}{\|\mathbf{v}\|}$$

$$\mathbf{0} \cdot \|\mathbf{v}\| = \lim_{t \to 0} (A\mathbf{v} - B\mathbf{v})$$

$$\mathbf{0} = A\mathbf{v} - B\mathbf{v}$$

$$B\mathbf{v} = A\mathbf{v}$$

- Example: Let $f: \mathbb{R}^n \to \mathbb{R}^m$ 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

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

- 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

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

- 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 $\mathbf{h} \to \mathbf{0}$, the right-hand side of the above equation goes to $f(\mathbf{a})$.
 - As a linear transformation, $A\mathbf{h} \to \mathbf{0}$ as $\mathbf{h} \to \mathbf{0}$.
 - Clearly $\|\mathbf{h}\| \to \mathbf{0}$ as $\mathbf{h} \to \mathbf{0}$.
 - And we have by definition that the last term goes to $\mathbf{0}$ as $\mathbf{h} \to \mathbf{0}$.
- Therefore, f is continuous at \mathbf{a} .
- Observation: A function $f: U \to \mathbb{R}^m$ is given by an m-tuple of functions $f_1: U \to \mathbb{R}$ 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

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

- 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 $\lim_{\mathbf{h}\to\mathbf{0}} L_i(\mathbf{h}) = \mathbf{0}$ iff the linear transformation $\mathbf{h}\mapsto (A\mathbf{h})_i:\mathbb{R}^m\to\mathbb{R}$ is the total derivative of f_i .
- Now, each f_i is a function of n variables, i.e., $f_i(x_1,\ldots,x_n)$ where x_1,\ldots,x_n are coordinates on \mathbb{R}^n .
- Partial derivative (of f wrt. x_i at $\mathbf{a} \in U$): The following quantity. Denoted by $\partial f/\partial x_i$. Given by

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

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

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- 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 $f: \mathbb{R}^2 \to \mathbb{R}$ 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

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

- 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,

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

• 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

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

- Jacobian matrix (of f at a): The above matrix, representing the total derivative of f at a.
- Theorem: Suppose $f: U \to \mathbb{R}^m$ 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 $g:[a,b]\to\mathbb{R}$ 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

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

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

Apply the MVT to each term to get

$$f(a_1,\ldots,a_i+h_i,\ldots,a_n+h_n)-f(a_1,\ldots,a_i,\ldots,a_n+h_n)=h_i\frac{\partial f}{\partial x_i}(a_1,\ldots,a_i+h_n)$$

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 a.
- WTS:

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

- 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 $\pi_i: \mathbb{R}^n \to \mathbb{R}^n$ 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

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

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

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

- ullet Note that this theorem gives a sufficient condition but not a necessary condition for f to be differentiable.
- - Note that $f: U \to \mathbb{R}^m$ 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

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

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

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

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

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

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

- For the second half of the fraction,

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

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

$$\frac{\left\|A\mathbf{h} + \tilde{f}(\mathbf{h})\right\|}{\|\mathbf{h}\|} \le \frac{\|A\mathbf{h}\|}{\|\mathbf{h}\|} + \frac{\left\|\tilde{f}(\mathbf{h})\right\|}{\|\mathbf{h}\|} \le \|A\| + \frac{\left\|\tilde{f}(\mathbf{h})\right\|}{\|\mathbf{h}\|}$$

where ||A|| is the operator norm and $||\tilde{f}(\mathbf{h})||/||\mathbf{h}|| \to 0$ as $\mathbf{h} \to \mathbf{0}$ 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 $f, g: U \to \mathbb{R}^m$ 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 $f, g: U \to \mathbb{R}$ 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 $f: U \to \mathbb{R}$ 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 $F: U \to \mathbb{R}^{2m}$ and $G: \mathbb{R}^m \times \mathbb{R}^m \to \mathbb{R}^m$ defined by

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

so that

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

- lacksquare 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

$$D(f+g)(\mathbf{a}) = D(G \circ F)(\mathbf{a})$$

$$= DG(F(\mathbf{a})) \circ DF(\mathbf{a})$$

$$= G(DF(\mathbf{a}))$$

$$= G(Df(\mathbf{a}), Dg(\mathbf{a}))$$

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

- Prove the others the same way.
- Theorem (Mean Value Theorem): Suppose $f: U \to \mathbb{R}$ 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 $Df: U \to \mathcal{L}(\mathbb{R}^n, \mathbb{R}^m)$.
 - 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} .
- \bullet We can ask for Df to itself be differentiable. We define

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

if it exists and, more generally,

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

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

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- Note that we technically need only require that $D^k f$ exist, as this implies the existence of $Df, \ldots, D^{k-1}f$.
- A function $f: U \to \mathbb{R}^m$ is of class C^k iff all partial derivatives $\partial f/\partial x_i: U \to \mathbb{R}^m$ 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^{∞} .
- Theorem: Let $U \subset \mathbb{R}^n$ be open, and let $f: U \to \mathbb{R}^m$ be a C^2 function. Then for any i, j,

$$\frac{\partial}{\partial x_i} \left(\frac{\partial f}{\partial x_j} \right) = \frac{\partial}{\partial x_j} \left(\frac{\partial f}{\partial x_i} \right)$$

- WLOG, let n=2 (because only two variables play a role in this theorem; we don't need the others) and m=1 (we can do this for each component separately).
- See the figure/proof associated with Theorem 15.3 in Labalme (2021).
- Note that there is a homework problem giving an example of a function for which the above quantities are defined but not equal. This doesn't violate the theorem, though, because the function isn't C^2 .
- Next goal: One of the most important theorems in this class the inverse function theorem.
- Theorem (Inverse function theorem): Suppose $U \subset \mathbb{R}^n$ is open, $f: U \to \mathbb{R}^n$ is a C^k function for $k \geq 1$, and $\mathbf{a} \in U$ such that $Df(\mathbf{a}): \mathbb{R}^n \to \mathbb{R}^n$ is invertible. Then there exist open neighborhoods $V \subset U$ of \mathbf{a} and W of $\mathbf{b} = f(\mathbf{a}) \in \mathbb{R}^n$ and a C^k function $g: W \to V$ such that f(V) = W, g(W) = V, $g(f(\mathbf{x})) = \mathbf{x}$ for all $\mathbf{x} \in V$, and $f(g(\mathbf{y})) = \mathbf{y}$ for all $\mathbf{y} \in W$. Moreover, $Dg(\mathbf{y}) = Df(g(\mathbf{y}))^{-1}$ for all $y \in W$.
 - The proof will be reasonably involved because we have to deal with the construction of the open neighborhoods. The proof will be reasonably nonconstructive?
 - Intuition: If n = 1, $f'(\mathbf{a}) \neq 0$ implies that f is increasing or decreasing in a neighborhood of \mathbf{a} ; in either case, there's a local inverse.
 - In higher dimensions, the derivative won't be up or down but will look like a linear transformation.
 - Idea of proof: Given $\mathbf{y} \in \mathbb{R}^n$ near $\mathbf{b} = f(\mathbf{a})$, we want to find $\mathbf{x} \in U$ near \mathbf{a} such that $f(\mathbf{x}) = \mathbf{y}$.
 - First guess: Take $\mathbf{x}_0 = \mathbf{a} A^{-1}(\mathbf{b} \mathbf{y})$ where $A = Df(\mathbf{a})$.
 - Hope: $f(\mathbf{x}_0)$ is closer to \mathbf{y} than $\mathbf{b} = f(\mathbf{a})$.
 - Then we iterate; do this again and again to get a sequence that converges to the point we want. In particular, $\mathbf{x}_1 = \mathbf{x}_0 A^{-1}(f(\mathbf{x}_0) \mathbf{y})$, and on and on.
 - We're going to formalize this idea of iteration using the **contraction mapping theorem** (aka the Banach fixed point theorem).
 - Let $\mathbf{y} \in \mathbb{R}^n$ be fixed near $\mathbf{b} = f(\mathbf{a})$. $F_{\mathbf{y}}(\mathbf{x}) = \mathbf{x} A^{-1}(f(\mathbf{x}) \mathbf{y})$. Note: $f(\mathbf{x}) = \mathbf{y}$ iff $F_{\mathbf{y}}(\mathbf{x}) = \mathbf{x}$.
 - Goal: Find a fixed point of $F_{\mathbf{y}}(\mathbf{x})$ (another way of iterating the sequence is that at some point you will get a fixed point, i.e., a point that you can plug into the recursion relation and get the same point back out).
- Theorem (Contraction mapping theorem): Let X be a complete metric space, and suppose $T: X \to X$ is a function for which there exists a constant r < 1 such that for all $\mathbf{x}, \mathbf{y} \in X$, $d(T(\mathbf{x}), T(\mathbf{y})) \le r \cdot d(\mathbf{x}, \mathbf{y})$. Then T has a unique fixed point.
- The way you find a fixed point is by starting with an arbitrary point and then just iterating T to it. Then show that in the limit it converges to something, and that something is a fixed point.
- You can't have more than one fixed point.

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} , $f: A \to \mathbb{R}^n$, and $\mathbf{u} \in \mathbb{R}^m$ is nonzero. Denoted by $f'(\mathbf{a}; \mathbf{u})$. Given by

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

- 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 **a** if $f'(\mathbf{a}; \mathbf{u})$ 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 $\phi: A \to \mathbb{R}$, where $A \subset \mathbb{R}$ contains a neighborhood of a, for which there exists a number λ such that

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

- **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 $f: A \to \mathbb{R}^n$. If f is differentiable at \mathbf{a} , then all the directional derivatives of f at \mathbf{a} exist, and

$$f'(\mathbf{a}; \mathbf{u}) = Df(\mathbf{a}) \cdot \mathbf{u}$$

- Theorem 5.2: Let $A \subset \mathbb{R}^m$, and let $f: A \to \mathbb{R}^n$. If f is differentiable at a, then f is continuous at 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 $D_i f(\mathbf{a})$.
- Theorem 5.3. Let $A \subset \mathbb{R}^m$, and let $f: A \to \mathbb{R}$. If f is differentiable at **a**, then

$$Df(\mathbf{a}) = \begin{bmatrix} D_1 f(\mathbf{a}) & \cdots & D_m f(\mathbf{a}) \end{bmatrix}$$

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

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

- (a) The function f is differentiable at \mathbf{a} if and only if each component function f_i is differentiable at \mathbf{a} .
- (b) 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 $D_j f_i(\mathbf{a})$.

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

$$Dg(\mathbf{x}) = \begin{bmatrix} D_1 g(\mathbf{x}) & D_2 g(\mathbf{x}) & D_3 g(\mathbf{x}) \end{bmatrix}$$

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 $\overset{\rightarrow}{\nabla} g$ and \mathbf{u} .
- Theorem 6.1 (Mean Value Theorem): If $\phi : [a,b] \to \mathbb{R}$ 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 $D_j f_i(\mathbf{x})$ 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 $D_j f_i(\mathbf{x})$ 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 $f: A \to \mathbb{R}$ 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 $A \subset \mathbb{R}^m$, $B \subset \mathbb{R}^n$, $f : A \to \mathbb{R}^n$, $g : B \to \mathbb{R}^p$, $f(A) \subset B$, 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.

¹See Theorem 15.3 in Labalme (2021).

- Corollary 7.2: Let A be open in \mathbb{R}^m , and let B be open in \mathbb{R}^n . Let $f: A \to \mathbb{R}^n$ and $g: B \to \mathbb{R}^p$ 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 $f: A \to \mathbb{R}$ 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 , $f: A \to \mathbb{R}^n$, 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 $i: \mathbb{R}^n \to \mathbb{R}^n$ 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

$$Dg(\mathbf{b}) \cdot Df(\mathbf{a}) = I_n$$

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

as desired. \Box

- It follows from Theorem 7.4 that for f^{-1} to be differentiable at **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).

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.