

LECTURE 11: MULTIVARIABLE ANALYSIS

Welcome to the heart of the course, where we generalize single-variable concepts like derivatives to multivariable calculus.

1. PRELIMINARIES

Notation: $x = (x_1, x_2, \dots, x_n)$ is a point in \mathbb{R}^n

The **length** of x is denoted by

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

We will frequently deal with linear mappings between spaces:

Definition: $A : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is a **linear transformation** if:

- (1) For all x and y , $A(x + y) = Ax + Ay$
- (2) For all x and all scalars c , $A(cx) = cAx$

The set of linear transformations from \mathbb{R}^n to \mathbb{R}^m is denoted as $\mathcal{L}(\mathbb{R}^n, \mathbb{R}^m)$

Mnemonic: $n = \mathbf{I}N$ put, $m = \mathbf{M}O$ uthput, it goes from \mathbb{R}^n to \mathbb{R}^m

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2. NORM OF A

Given a linear transformation A , we can define $\|A\|$, the norm of A , which like a “maximum spread” of A :

Definition: If $A \in \mathcal{L}(\mathbb{R}^n, \mathbb{R}^m)$ then

$$\|A\| = \sup_{x \in \mathbb{R}^n} \frac{|Ax|}{|x|}$$

Note: By definition of sup we have $\|A\| \geq \frac{|Ax|}{|x|}$ for all x and hence

$$|Ax| \leq \|A\| |x|$$

This is the form we’ll most frequently use.

Intuitively: Think of $\|A\|$ as the largest possible spread of A . For example, if $\|A\| = 2$ then we have $|Ax| \leq 2|x|$ for all x , so $|Ax|$ is never more than twice as big as $|x|$.

$\|A\| = \infty$ concretely means there is a sequence x_k of points such that

$$\lim_{k \rightarrow \infty} \frac{|Ax_k|}{|x_k|} = \infty$$

So the stretch $\frac{|Ax|}{|x|}$ gets uncontrollably big.

Note: In the definition of $\|A\|$, it’s actually enough to assume $|x| \leq 1$ or even just $|x| = 1$ because

$$\frac{|A(cx)|}{|cx|} = \frac{|cAx|}{|c||x|} = \frac{|c||Ax|}{|c||x|} = \frac{|Ax|}{|x|}$$

So if the property holds for x it holds for any multiple cx .

3. PROPERTIES OF $\|A\|$

Let's now prove some properties of $\|A\|$.

First of all, could $\|A\| = \infty$? Not in finite dimensions!

Fact: If $A \in \mathcal{L}(\mathbb{R}^n, \mathbb{R}^m)$ then $\|A\| < \infty$ (that is A is **bounded**)

Proof: Finite-dimensionality is crucial here!

Let $\{e_1, \dots, e_n\}$ be the standard basis of \mathbb{R}^n and let $x \in \mathbb{R}^n$ be given. Then there are c_1, \dots, c_n such that

$$x = c_1 e_1 + \dots + c_n e_n$$

$$\text{Then } |Ax| = \left| A \left(\sum_{i=1}^n c_i e_i \right) \right| \stackrel{\text{LIN}}{=} \left| \sum_{i=1}^n c_i (Ae_i) \right| \leq \sum_{i=1}^n |c_i| |Ae_i|$$

However, for each i , we have

$$|c_i| = \sqrt{(c_i)^2} \leq \sqrt{(c_1)^2 + \dots + (c_n)^2} = |x|$$

$$\text{Hence } |Ax| \leq \sum_{i=1}^n \underbrace{|c_i|}_{\leq |x|} |Ae_i| \leq \left(\sum_{i=1}^n |Ae_i| \right) |x| = C |x|$$

Where $C =: \sum_{i=1}^n |Ae_i|$ (doesn't depend on x).

From this it follows that for all x , $\frac{|Ax|}{|x|} \leq C < \infty$

□

One of the most surprising facts about linear transformations is that A bounded implies A is continuous!

Fact: If $A \in \mathcal{L}(\mathbb{R}^n, \mathbb{R}^m)$ then A is uniformly continuous

Proof: Let $\epsilon > 0$ be given, let $\delta = \frac{\epsilon}{\|A\|}$ then if $|x - y| < \delta$ then

$$|Ax - Ay| = |A(x - y)| \leq \|A\| |x - y| < \|A\| \left(\frac{\epsilon}{\|A\|} \right) = \epsilon \checkmark \quad \square$$

Next we want to show that $\|A\|$ behaves nicely with respect to sums and compositions

Facts:

$$(1) \|A + B\| \leq \|A\| + \|B\|$$

$$(2) \|AB\| \leq \|A\| \|B\|$$

Here AB is the composition of A and B whenever it's defined

Why? (1) follows because for all x

$$|(A + B)x| = |Ax + Bx| \leq |Ax| + |Bx| \leq \|A\| |x| + \|B\| |x| = (\|A\| + \|B\|) |x|$$

And (2) follows similarly because for all x

$$|(AB)x| = |A(Bx)| \leq \|A\| |Bx| \leq \|A\| \|B\| |x|$$

Fun fact: If you define $d(A, B) = \|A - B\|$, then $\mathcal{L}(\mathbb{R}^n, \mathbb{R}^m)$ actually becomes a metric space! So all the concepts of open sets and continuity makes sense for linear transformations as well. We'll discuss this below.

Relation to matrices: If A is a linear transformation with matrix

$$[A] = \begin{bmatrix} a_{11} & \cdots & a_{1n} \\ \vdots & a_{ij} & \vdots \\ a_{m1} & \cdots & a_{mn} \end{bmatrix}$$

Then by calculating $|Ax|$ and using Cauchy-Schwarz, you can actually show that (see section 9.9 of Rudin for details)

$$\|A\| \leq \sqrt{\sum_{i=1}^m \sum_{j=1}^n (a_{ij})^2}$$

There are examples where we get a strict inequality. For example, if

$$A(x_1, x_2) = (x_1, 2x_2)$$

Then can show $\|A\| = 2$ but

$$[A] = \begin{bmatrix} 1 & 0 \\ 0 & 2 \end{bmatrix}$$

And so the square root of the sum of squares of components is $\sqrt{5}$

4. A^{-1}

Later in this chapter, we will talk a lot about inverse transformations. In order to do this, let's study the inverse transformation A^{-1} a bit more carefully.

Notation: $\mathcal{L}(\mathbb{R}^n) = \mathcal{L}(\mathbb{R}^n, \mathbb{R}^n)$ (space of linear operators on \mathbb{R}^n)

Recall: $A \in \mathcal{L}(\mathbb{R}^n)$ is **invertible** if there is $A^{-1} \in \mathcal{L}(\mathbb{R}^n)$ such that $AA^{-1} = A^{-1}A = I$ (the identity transformation)

Notation: Ω = set of invertible linear transformations on \mathbb{R}^n

Goal: Show Ω is open and moreover the mapping $A \rightarrow A^{-1}$ is continuous (with respect to the metric $d(A, B) = \|B - A\|$)

Lemma: If $A \in \Omega$ and $B \in \mathcal{L}(\mathbb{R}^n)$ is such that

$$\|B - A\| \|A^{-1}\| < 1$$

Then $B \in \Omega$

So if B is close enough to A , then B is invertible as well. This makes sense *if* Ω were open, and in fact helps us to show open-ness (see below)

Proof:

STEP 1: Let $\alpha = \frac{1}{\|A^{-1}\|}$ and $\beta = \|B - A\|$

$$\text{Then } \beta = \|B - A\| < \frac{1}{\|A^{-1}\|} = \alpha \Rightarrow \beta < \alpha$$

Then for every $x \in \mathbb{R}^n$, consider:

$$\begin{aligned} \alpha |x| &= \alpha |A^{-1}Ax| \leq \alpha \underbrace{\|A^{-1}\|}_{\frac{1}{\alpha}} |Ax| = |Ax| \\ &= |(A - B)x + Bx| \\ &\leq |(A - B)x| + |Bx| \\ &\leq \|A - B\| |x| + |Bx| \\ &= \beta |x| + |Bx| \end{aligned}$$

Therefore we obtain

$$(\alpha - \beta) |x| \leq |Bx|$$

STEP 2: Claim: B is one-to-one

If $Bx = 0$ then since $\alpha > \beta$ we get

$$\underbrace{(\alpha - \beta)}_{>0} |x| \leq |Bx| = |0| = 0$$

Which in turn implies $|x| = 0$ and so $x = 0$ ✓

Since $B : \mathbb{R}^n \rightarrow \mathbb{R}^n$ is one-to-one, it follows that B is also onto and hence B is invertible, that is $B \in \Omega$ \square

Corollary: Ω is open

Suppose $A \in \Omega$ and let $r = \frac{1}{\|A^{-1}\|} > 0$ then the previous result shows that if $d(B, A) < r$ then $B \in \Omega$, so Ω is open. \square

Fact: The mapping $A \rightarrow A^{-1}$ is continuous

STEP 1: We need to study $B^{-1} - A^{-1}$ for B “close” to A

Claim # 1:

$$B^{-1} - A^{-1} = B^{-1}(A - B)A^{-1}$$

Why?

$$\begin{aligned} B^{-1}(A - B)A^{-1} &= B^{-1}(AA^{-1} - BA^{-1}) = B^{-1}(I - BA^{-1}) = B^{-1} - B^{-1}BA^{-1} \\ &= B^{-1} - A^{-1} \checkmark \end{aligned}$$

From the claim, it follows that

$$\|B^{-1} - A^{-1}\| \leq \|B^{-1}\| \underbrace{\|A - B\|}_{\beta} \underbrace{\|A^{-1}\|}_{\frac{1}{\alpha}} = \|B^{-1}\| \left(\frac{\beta}{\alpha} \right)$$

So all is left is to study $\|B^{-1}\|$

STEP 2:

Claim # 2:

$$\|B^{-1}\| \leq \frac{1}{\alpha - \beta}$$

Why? In the Lemma, we showed that for all x , we have

$$(\alpha - \beta) |x| \leq |Bx|$$

Replacing x with $B^{-1}x$ in the above, we get

$$(\alpha - \beta) |B^{-1}x| \leq |BB^{-1}x| \Rightarrow (\alpha - \beta) |B^{-1}x| < |x|$$

And therefore $\frac{|B^{-1}x|}{|x|} < \frac{1}{\alpha - \beta}$ and hence $\|B^{-1}\| \leq \frac{1}{\alpha - \beta}$

STEP 3: Combining steps 1 and 2 we get

$$\|B^{-1} - A^{-1}\| \leq \|B^{-1}\| \left(\frac{\beta}{\alpha} \right) \leq \frac{\beta}{(\alpha - \beta)\alpha}$$

Let $\epsilon > 0$ be given, then since

$$\lim_{\beta \rightarrow 0} \frac{\beta}{(\alpha - \beta)\alpha} = 0$$

There is $\delta > 0$ such that if $|\beta| < \delta$ then $\left| \frac{\beta}{(\alpha - \beta)\alpha} \right| < \epsilon$.

With that δ , if $\underbrace{\|A - B\|}_{\beta} < \delta$ then by the above, $\|B^{-1} - A^{-1}\| < \epsilon$. \square

5. THE DERIVATIVE IN \mathbb{R}^n

With those preliminaries out of our way, we can finally embark on our exploration of derivatives in \mathbb{R}^n .

Goal: If $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$, how to define the derivative $f'(x)$?

First guess: By analogy with the scalar case, if $x \in \mathbb{R}^n$

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

But here h is a vector, so it makes no sense to divide by h

So we need a definition of derivatives that doesn't have division in it.

What saves us is the concept of linear approximation from calculus:

Analogy: ($n = 1$) Note that if h is small, then

$$f(x+h) = f(x) + f'(x)h + \text{Smaller terms}$$

(This was used to approximate quantities like $\sqrt{4.02}$ or $\ln(0.97)$)

This is the point of view that we'll take:

Important definition: Suppose $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$ and $x \in \mathbb{R}^n$.

If there is a linear transformation $A : \mathbb{R}^n \rightarrow \mathbb{R}^m$ such that

$$f(x+h) = f(x) + Ah + r(h)$$

$$\text{Where } \lim_{h \rightarrow 0} \frac{|r(h)|}{|h|} = 0$$

Then we say f is **differentiable at** x and $f'(x) = A$

Definition: f is **differentiable** if f is differentiable at all x

In other words, if you can expand $f(x+h)$ out with a small remainder, then the linear part is the derivative of f .

Before, $f'(x)$ was just a number, but now it's something more dynamic, it's a linear transformation. Intuitively, if f distorts space, then $f'(x)$ describes the linear part of the distortion.

Note: More commonly, people write $o(h)$ instead of $r(h)$, it's a term smaller than h