# Multivariable Calculus Notes

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## Chapter 1

## Sequences

We know the usual definition of sequences in  $\mathbb{R}$ . A sequence of vectors converges similarly, except that the norm used in the definition of convergence is now a vector norm, not simply absolute value. A sequence of vectors  $\{\vec{a}_i\}$  in  $\mathbb{R}^n$  converges to  $\vec{a} \in \mathbb{R}^n$  if for any  $\epsilon > 0$  there exists a  $n \in \mathbb{N}$  such that

$$n > N \implies \|\vec{a}_i - \vec{a}\| < \epsilon$$

This definition can be related to the convergence of the vector components.

**Prop 1.1.** Let  $\{\vec{a}_i\}$  be a sequence of vectors (coordinate vectors) in  $\mathbb{R}^n$ . Then  $\{\vec{a}_i\}$  converges to  $\vec{a} \in \mathbb{R}^n$  if and only if each component  $(\vec{a}_i)_i$  converges to  $(\vec{a})_i$ .

*Proof.* Assume that we have the components converging. That is, for  $\epsilon > 0$  we have some  $N \in \mathbb{N}$  such that

$$n > N \implies \left| (\vec{a}_i)_j - (\vec{a})_j \right| < \epsilon / \sqrt{n}$$

We choose N to hold for every component i. This can be arranged by taking N to the the maximum of the N guaranteed for each individual component. Thus, we have, for n > N,

$$\|\vec{a}_i - \vec{a}\|^2 = \sum_{j=1}^n \left| (\vec{a}_i)_j - (\vec{a})_j \right|^2$$

$$< \sum_{j=1}^n \epsilon^2 / n$$

$$= \epsilon$$

so that  $\{\vec{a}_i\}$  converges to  $\vec{a}$ .

Conversely, assume that  $\{\vec{a}_i\}$  converges to  $\vec{a}$ . Let  $\epsilon > 0$  be arbitrary and let N be the guaranteed natural number in the definition of vector sequence convergence.

The key is to note that for any j = 1, 2, ..., n we have

$$\left| (\vec{a}_i)_j - (\vec{a})_j \right|^2 \le \sum_{k=1}^n \left| (\vec{a}_i)_k - (\vec{a})_k \right|^2 < \epsilon^2$$

whenever n > N. Thus, we have in particular

$$n > N \implies \left| \left( \vec{a}_i \right)_j - \left( \vec{a} \right)_j \right| < \epsilon$$

and so the component sequences converge as expected.

Let's look at some interesting consequences.

**Prop 1.2.** Let  $\vec{x}_k \to \vec{a}$ . Then  $||\vec{x}_k|| \to ||\vec{a}||$  and  $\vec{b} \cdot \vec{x}_k \to \vec{b} \cdot \vec{a}$ .

*Proof.* The first can be proven by writing  $\vec{x} = [x_1, x_2, \dots, x_n]$  and proceeding as usual with the usual 2-norm. More generally, though, we can use the (reverse) triangle inequality property of the norm to see that

$$|||\vec{x}_k|| - ||\vec{a}||| \le ||\vec{x}_k - \vec{a}||$$

which can be made smaller than arbitrary  $\epsilon > 0$  by proper choice of N. Thus, the sequence of norms converges as expected.

Next, we could let  $f(\vec{x}) = b^T \vec{x}$  and use continuity of linear functions to prove the conclusion. Instead, just notice that, from a useful proposition in the next chapter,

$$\left\| \vec{b}^T \vec{x} \right\| \le \left\| \vec{b} \right\| \left\| \vec{x} \right\|$$

$$\left\|\vec{b}^T\vec{x}_k - \vec{b}^T\vec{a}\right\| = \left\|\vec{b}^T\left(\vec{x}_k - \vec{a}\right)\right\| = \left\|\vec{b}\right\| \left\|\vec{x}_k - \vec{a}\right\|$$

so that we have  $\|\vec{b}^T\vec{x}_k - \vec{b}^T\vec{a}\| = \|\vec{b}^T(\vec{x}_k - \vec{a})\| = \|\vec{b}\| \|\vec{x}_k - \vec{a}\|$  Since  $\vec{x}_k \to \vec{a}$ , given arbitrary  $\epsilon > 0$ , there is some N such that  $k > N \implies \|\vec{x}_k - \vec{a}\| < \epsilon / \|\vec{b}\|$ . Thus, we have proven the statement.

#### Subsequences 1.1

A subsequence of a sequence  $\{\vec{x}_k\}_k$  is a subset  $\{\vec{x}_{k_i}\}_{k_i}$  where  $k_1 < k_2 < k_3 < \dots$  is an increasing sequence of natural numbers.

**Prop 1.3.** If a sequence converges, then any subsequence converges. If every subsequence of a sequence converges, then the sequence converges.

*Proof.* The converse is trivial, since the sequence itself is a subsequence. Note that it is not the case that a single convergent subsequence implies convergence of the

Let  $\{\vec{x}_{k_i}\}_{k_i}$  be a subsequence of convergent sequence  $\{\vec{x}_k\}_k$  which has limit  $\vec{a}$ .

Then for arbitrary  $k>N \implies \|\vec{x}_k-a\| < \epsilon$  Since  $\{\vec{x}_{k_i}\}_{k_i}$  is a subsequence, we then have  $i>N \implies \|\vec{x}_{k_i}-\vec{a}\| < \epsilon$ 

$$k > N \implies \|\vec{x}_k - \vec{a}\| < \epsilon$$

$$i > N \implies \|\vec{x}_{k:} - \vec{a}\| < \epsilon$$

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# Chapter 2

# Continuity

**Prop 2.1.** Let  $T: \mathbb{R}^n \to \mathbb{R}^m$  be a linear map represented by the matrix  $A \in \mathbb{R}^{m \times n}$ . Then we have

$$||T(h)|| = ||Ah|| \le M ||h||$$

for some constant M.

*Proof.* Note that this is just a statement about matrix-vector multiplication. Let  $[Ah]_i$  denote the *i*th entry of the matrix product  $Ah \in \mathbb{R}^m$ . That is,

$$[Ah]_i = \sum_{j=1}^n A_{ij} h_j$$

Then let

$$r = \max_{i,j} |A_{ij}|$$

$$||h||^2 = \sum_{j=1}^n |h_j|^2$$

we have

$$|h_j| \le ||h||$$

Taken together, we have

$$\left[\left.Ah\right]_{i} = \sum_{j=1}^{n} A_{ij}h_{j} \leq \sum_{j=1}^{n} \left|A_{ij}\right| \left|h_{j}\right| \leq rn \left\|h\right\|$$

so that

so that 
$$\|Ah\|^2=\sum_{i=1}^m \left[Ah\right]_i^2 \leq \sum_{i=1}^m r^2n^2 \|h\|^2=mr^2n^2 \|h\|^2$$
 and so we have 
$$\|Ah\| \leq rn\sqrt{m} \, \|h\|$$
 Letting  $M=rn\sqrt{m},$  the conclusion holds.

$$||Ah|| \le rn\sqrt{m} ||h||$$

**Lemma 2.2.** Let  $T: \mathbb{R}^n \to \mathbb{R}^m$  be a linear map represented by the matrix  $A \in \mathbb{R}^{m \times n}$ . Then for  $a \in \mathbb{R}^n$ , T is continuous at a.

$$\lim_{h \to 0} ||T(a+h) - T(a)|| = 0$$

*Proof.* We must show that 
$$\lim_{h\to 0}\|T\left(a+h\right)-T\left(a\right)\|=0$$
 Note that, by linearity, 
$$\|T\left(a+h\right)-T\left(a\right)\|=\|T\left(a\right)+T\left(h\right)-T\left(a\right)\|=\|T\left(h\right)\|\leq M\left\|h\right\|$$
 and the right-hand side clearly goes to 0 as  $h\to 0$ .

The next seems to be an alternative approach. For arbitrary linear map  $A \in \mathbb{R}^{m \times n}$ , we can define a matrix norm

$$||A|| \coloneqq \sum_{i=1,\dots,m} \sum_{j=1,\dots,n} A_{i,j}^2$$

This is technically the square of the element-wise matrix p-norm:

$$\|A\|_p = \left(\sum_{\text{rows cols}} \sum_{i,j} A_{i,j}^p\right)^{1/p}$$

Then we can look at bounds on matrix-vector and matrix-matrix products. In fact, the following proposition is a generalization of the one above.

**Prop 2.3.** Let  $A \in \mathbb{R}^{m \times n}$ ,  $B \in \mathbb{R}^{n \times p}$ , and  $\vec{b} \in \mathbb{R}^n$ . Then

$$||A\vec{b}|| \le ||A|| \, ||\vec{b}||$$

$$\|AB\| \leq \|A\| \, \|B\|$$

*Proof.* For the first inequality, notice that the magnitude on the left is a vector norm.

$$\left\| A\vec{b} \right\|^2 = \sum_{i=1}^m \left( A\vec{b} \right)_i^2$$

and that each entry in this vector is the inner product (usual Euclidean dot product)

$$(A\vec{b})_i = A_{i*}\vec{b}$$

Since this is an inner product, we may apply the Cauchy-Schwarz inequality to it

$$(A\vec{b})_{i} = A_{i*}\vec{b} \le |A_{i*}\vec{b}| \le |A_{i*}| \|\vec{b}\|$$

Taking this in the original vector magnitude, we have

$$\|A\vec{b}\|^{2} = \sum_{i=1}^{m} (A\vec{b})_{i}^{2}$$

$$\leq \sum_{i=1}^{m} \|A_{i*}\|^{2} \|\vec{b}\|^{2}$$

$$= \left(\sum_{i=1}^{m} \|A_{i*}\|^{2}\right) \|\vec{b}\|^{2}$$

$$= \|A\|^{2} \|\vec{b}\|^{2}$$

and so  $||A\vec{b}|| \le ||A|| ||\vec{b}||$  as we wanted to prove.

The next inequality is even more general and follows from the matrix-vector product case. Note that every column of the matrix product AB is of the form  $A\vec{b}$  where  $\vec{b} = B_{*j}$ . Thus

$$||AB||^{2} = \sum_{i=1}^{m} \sum_{j=1}^{n} (AB)_{i,j}^{2}$$

$$= \sum_{j=1}^{n} |AB_{*j}|^{2}$$

$$\leq \sum_{j=1}^{n} ||A||^{2} ||B_{*j}||^{2}$$

$$= ||A||^{2} \left( \sum_{j=1}^{n} ||B_{*j}||^{2} \right)$$

$$= ||A||^{2} ||B||^{2}$$

and so

$$\|AB\| \le \|A\| \|B\|$$

as desired.

The second inequality defines the matrix norm  $\|\cdot\|$  (and more generally the matrix 2-norm) as a *submultiplicative* norm. With these general properties in hand, we can easily verify continuity of a linear function.

*Proof.* The linear function  $T: \mathbb{R}^n \to \mathbb{R}^m$  will be continuous if we have

$$\lim_{x\to a}T\left(x\right)=T\left(a\right)$$

Let  $\epsilon > 0$  be arbitrary. Then

$$|\vec{x} - \vec{a}| < \epsilon / \|A\|$$

implies that

$$||T(\vec{x}) - T(\vec{a})|| = ||A\vec{x} - A\vec{a}||$$

$$= ||A(\vec{x} - \vec{a})||$$

$$\leq ||A|| ||\vec{x} - \vec{a}||$$

$$\leq \epsilon$$

Note that linearity of T allowed us to express its action as a matrix multiplication, and the associative (linearity) property of matrix multiplication allowed us to "pull it off" of the  $\vec{x}$  and  $\vec{a}$  and onto their vector difference. Then, the previous proposition allowed us to bound this matrix-vector product by the norm of the vector and a quantity which measures in some sense the size of the matrix A.

### Chapter 3

## The Derivative

Let

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

be an arbitrary function. Then the derivative of f at  $a \in \mathbb{R}^n$  is the unique linear map

$$\lambda: \mathbb{R}^n \to \mathbb{R}^m$$

such that

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

The norms are required since all quantities are vectors. Note that since  $\lambda$  is a linear map, it has a matrix representation, written Df(a) and called the derivative of f at a. The action of  $\lambda$  on h is

$$\lambda\left(h\right) = Df\left(a\right)h$$

i.e., the usual matrix multiplication.

**Prop 3.1.** If  $f: \mathbb{R}^n \to \mathbb{R}^m$  is differentiable at  $a \in \mathbb{R}^n$ , then this derivative is unique.

*Proof.* Let  $f: \mathbb{R}^n \to \mathbb{R}^m$  be differentiable at  $a \in \mathbb{R}^n$  and assume there are two derivatives  $\lambda$  and  $\mu$ . Then  $\|\lambda(h) - \mu(h)\| \le \|\lambda(h) - f(a+h) + f(a)\| + \|\mu(h) - f(a+h) + f(a)\|$  and upon dividing by  $\|h\|$  and taking the limit  $h \to 0$ , we obtain

$$\|\lambda(h) - \mu(h)\| \le \|\lambda(h) - f(a+h) + f(a)\| + \|\mu(h) - f(a+h) + f(a)\|$$

$$\lim_{h\to 0}\frac{\|\lambda\left(h\right)-\mu\left(h\right)\|}{\|h\|}=0$$

Choosing arbitrary  $x \in \mathbb{R}^n \setminus \{0\}$ , note that  $tx \to 0$  when  $t \to 0$   $(t \in \mathbb{R})$ . Then

$$0 = \lim_{t \to 0} \frac{\|\lambda(tx) - \mu(tx)\|}{\|tx\|} = \lim_{t \to 0} \frac{|t| \|\lambda(x) - \mu(x)\|}{|t| \|x\|} = \frac{\|\lambda(x) - \mu(x)\|}{\|x\|}$$

and so we have  $\lambda(x) = \mu(x)$  for arbitrary  $x \in \mathbb{R}^n$ . Thus, the derivative is unique.

**Prop 3.2.** If  $f: \mathbb{R}^n \to \mathbb{R}^m$  is differentiable at  $a \in \mathbb{R}^n$ , then it is continuous at a.

$$||f(a+h)-f(a)|| \le ||f(a+h)-f(a)-\lambda(h)|| + ||\lambda(h)||$$

$$||f(a+h) - f(a)|| \le ||f(a+h) - f(a) - \lambda(h)|| + ||\lambda(h)||$$
 so that 
$$||f(a+h) - f(a)|| \le ||h|| \frac{||f(a+h) - f(a) - \lambda(h)||}{||h||} + ||\lambda(h)||$$

By taking the limit of both sides as  $h \to 0$ , the first term on the right-hand side is 0, and the second term, containing the linear  $\lambda$ , goes to 0 as well. Specifically, from a property of linear maps proved in the previous chapter, we have

$$\|\lambda(h)\| \leq M\|h\|$$

for some constant M which certainly tends toward 0. Specifically, we have shown

$$\|\lambda(h)\| \le \|Df(a)\| \|h\|$$

where  $\|Df(a)\|$  is the squared entry-wise matrix norm of the derivative (matrix) of f at a, which certainly exists (is finite). Thus, we have

$$\lim_{h\to 0}\|f\left(a+h\right)-f\left(a\right)\|=0,$$
 establishing continuity of  $f$  at  
  $a.$