MA 200: Multivariable Calculus

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The course

Grading

• Homework: 20%

• Quizzes: 20%

• Midterm: 20%

• Final: 40%

Textbooks

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Lecture 1. Friday
August 2

Chapter 1

Linear algebra

Definition 1.1 (homogeneous function). Let V be a vector space over \mathbb{R} . A function $f: V \setminus \{0\} \to \mathbb{R}$ is called a homogeneous function of degree k if

$$f(rx) = r^k f(x)$$

for each $x \in V \setminus \{0\}$ and r > 0.

Remarks.

- If f and g are homogeneous functions of degree k and l respectively, then $f \cdot g$ is homogeneous of degree k + l and f/g is homogeneous of degree k l (provided g is never zero).
- $f \equiv 0$ is homogeneous of any degree.

Definition 1.2 (norm). Let V be a vector space over \mathbb{R} . A norm $\|\cdot\|$ on V is a function from V to \mathbb{R} that satisfies

- (N1) (Positivity) $||x|| \ge 0$ for any $x \in V$.
- (N2) (Definiteness) ||x|| = 0 iff x = 0.
- (N3) (Homogeneity) ||rx|| = |r|||x|| for any $x \in V$ and $r \in \mathbb{R}$.
- (N4) (Triangle inequality) $||x + y|| \le ||x|| + ||y||$ for any $x, y \in V$.

Definition 1.3 (normed linear space). A vector space V equipped with a norm $\|\cdot\|$ is called a *normed linear space*.

Remark. Any normed linear space $(V, \|\cdot\|)$ can be given a metric space structure by defining the distance d(x, y) between $x, y \in V$ as $\|x - y\|$.

The set $B(x,r) := \{y \in V \mid ||x-y|| < r\}$ is called the open ball of radius r centered at x.

The set $S(x,r) \coloneqq \{y \in V \mid ||x-y|| = r\}$ is called the sphere of radius r centered at x.

Exercise 1.4 (reverse triangle inequality). Let V be a normed linear space. Show that

$$|||x|| - ||y||| \le ||x - y|| \tag{1.1}$$

for any $x, y \in V$.

Proof. First observe from homogeneity (N3) that ||x|| = ||-x|| for any $x \in V$. Next, from the triangle inequality (N4) we have

$$||x|| \le ||x - y|| + ||y||$$

so that

$$||x|| - ||y|| \le ||x - y||.$$

Similarly,

$$||y|| \le ||y - x|| + ||x||$$

so that

$$-\|x - y\| \le \|x\| - \|y\|.$$

Combining these gives the result.

This shows that $f = x \mapsto ||x||$ is a (Lipschitz) continuous function on V.

Definition 1.5 (metric space). A *metric space* is a set X equipped with a function $d: X \times X \to \mathbb{R}$ called a *metric* that satisfies the following properties:

- (M1) $d(x,y) \ge 0$ for any $x,y \in X$.
- (M2) d(x, y) = 0 iff x = y.
- (M3) d(x,y) = d(y,x) for any $x, y \in X$.
- (M4) $d(x,z) \le d(x,y) + d(y,z)$ for any $x,y,z \in X$.

Exercise 1.6 (self). Show that any normed linear space $(V, \|\cdot\|)$ is a metric space under the distance $d(x, y) = \|x - y\|$.

Proof. (M1) and (M2) are immediate from (N1) and (N2). (N3) implies (M3) by scaling by -1. Triangle implies triangle.

Definition 1.7 (continuity). Let (X,d) and (Y,ρ) be metric spaces. A function $f: X \to Y$ is called *continuous* at $a \in X$ iff

$$x_n \to a \implies f(x_n) \to f(a)$$
, or $d(x_n, a) \to 0 \implies \rho(f(x_n), f(a)) \to 0$

Exercise 1.8 (product metric spaces). Let (X_1, d_1) and (X_2, d_2) be metric spaces. Let $d: X_1 \times X_2 \to \mathbb{R}$ be defined by

$$d((x_1, x_2), (y_1, y_2)) := d_1(x_1, y_1) + d_2(x_2, y_2).$$

Show that d is a metric on $X_1 \times X_2$.

Let $(z_n)_{n\in\mathbb{N}} = ((x_n, y_n))_{n\in\mathbb{N}}$ be a sequence in $X_1 \times X_2$. Show that $z_n \to (x, y)$ iff $x_n \to x$ and $y_n \to y$.

Proof. Suppose $x_n \to x$ and $y_n \to y$. That is, $d_1(x_n, x) \to 0$ and $d_2(y_n, y) \to 0$. Thus $d_1(x_n, x) + d_2(y_n, y) \to 0$.

Conversely if $d_1(x_n, x) + d_2(y_n, y) \to 0$ and each is nonnegative, then $d_1(x_n, x) \to 0$ and $d_2(y_n, y) \to 0$.

Remark. \tilde{d} given by

$$\widetilde{d}((x_1, x_2), (y_1, y_2)) := \min\{d_1(x_1, y_1), d_2(x_2, y_2)\}$$

is not a metric on $X_1 \times X_2$ as it fails definiteness.

However, $\max\{d_1, d_2\}$ is a metric.

Exercise 1.9. Let $(V, \|\cdot\|)$ be a normed linear space.

- The addition map $(x,y) \mapsto x + y$ is a continuous map from $V \times V$ to V.
- The scalar multiplication map $(\alpha, x) \mapsto \alpha x$ is continuous from $\mathbb{R} \times V$ to V.

Solution.

- $||x' + y' (x + y)|| \le ||x' x|| + ||y' y|| = ||(x', y') (x, y)||$.
- $\|\alpha'x' \alpha x\| \le \|\alpha'x' \alpha x'\| + \|\alpha x' \alpha x\| = |\alpha' \alpha|\|x'\| + |\alpha|\|x' x\|$. Thus choosing $\delta = \varepsilon / \max\{|\alpha|, \|x\|\}$ gives

$$\|\alpha'x' - \alpha x\| \le \max\{|\alpha|, \|x\|\}(|\alpha' - \alpha| + \|x' - x\|) < \varepsilon$$

whenever $|\alpha' - \alpha| + ||x' - x|| < \delta$.

Repeated in problem 1.1.

Examples.

• $(\ell^p \text{ norm}) \mathbb{R}^n \text{ with } p \in [1, \infty] \text{ and }$

$$||x||_p := (|x_1|^p + \dots + |x_n|^p)^{1/p}$$

where

$$||x||_{\infty} \coloneqq \max\{|x_1|,\ldots,|x_n|\}$$

is the limit of the l^p norms as $p \to \infty$.

Exercise 1.10. See problem 1.6.

Definition 1.11 (norm equivalence). Let $\|\cdot\|_a$ and $\|\cdot\|_b$ be two norms on V. We say that $\|\cdot\|_a$ and $\|\cdot\|_b$ are *equivalent* if these exist $c_1, c_2 > 0$ such that

$$c_1 ||x||_a \le ||x||_b \le c_2 ||x||_a$$

for all $x \in V$. We write $\|\cdot\|_a \sim \|\cdot\|_b$.

Exercise 1.12. Check that \sim is an equivalence relation.

Solution. Reflexivity is obvious. Symmetry is since

$$c_1 ||x||_a \le ||x||_b \le c_2 ||x||_a \implies \frac{1}{c_2} ||x||_b \le ||x||_a \le \frac{1}{c_1} ||x||_b.$$

For transitivity, let

$$c_1 ||x||_a \le ||x||_b \le c_2 ||x||_a,$$

$$c_3 ||x||_b \le ||x||_c \le c_4 ||x||_b.$$

Then

$$c_1 c_3 ||x||_a \le ||x||_c \le c_2 c_4 ||x||_a.$$

Lecture 2.

let Monday

en August 5

Proposition 1.13. Equivalent norms induce the same topology. That is, let $\|\cdot\|_a \sim \|\cdot\|_b$. Then a set is open (resp. compact) under $\|\cdot\|_a$ iff it is open (resp. compact) under $\|\cdot\|_b$.

Proof. Suppose $c_1 ||x||_a \le ||x||_b \le c_2 ||x||_a$.

Let $U \subseteq V$ be open under $\|\cdot\|_a$. Let $x \in U$. There exists $\varepsilon > 0$ such that $\|y - x\|_a < \varepsilon \implies y \in U$. But then $\|y - x\|_b < c_1 \varepsilon \implies y \in U$. Thus U is open under $\|\cdot\|_b$.

Compactness follows from openness.

Proposition 1.14. Every ℓ^p norm is equivalent to ℓ^{∞} .

Proof. Let
$$x \in \mathbb{R}^n$$
. Then $||x||_{\infty} \le ||x||_p \le n^{\frac{1}{p}} ||x||_{\infty}$.

The usual topology on \mathbb{R}^n is the one induced by the Euclidean norm. This norm itself is induced by the inner product $\langle x, y \rangle = \sum_{i=1}^n x_i y_i$. Using Cauchy-Schwarz, we can define the angle between two vectors $x, y \in \mathbb{R}^n$ to be

$$\cos^{-1}\left(\frac{\langle x, y\rangle}{\|x\|\|y\|}\right).$$

Proposition 1.15. Let $\|\cdot\|$ be any norm on \mathbb{R}^n . Then the function $x \mapsto \|x\|$ is Lipschitz continuous with respect to the Euclidean topology.

Proof.

$$||x|| = \left\| \sum x_i e_i \right\|$$

$$\leq \sum |x_i| ||e_i||$$

$$\leq M||x||_2$$

where $M = \sum ||e_i||$.

The reverse triangle inequality gives

$$|||x|| - ||y||| \le ||x - y||$$

$$\le M||x - y||_2.$$

Theorem 1.16. Any two norms on \mathbb{R}^n are equivalent.

Proof. Let $\|\cdot\|$ be any norm on \mathbb{R}^n . Then $x \mapsto \|x\|$ is continuous with respect to $\|\cdot\|_2$. Let

$$S(0,1)_{\|\cdot\|_2} = \{x \in \mathbb{R}^n : \|x\|_2 = 1\} = S^{n-1}.$$

 $\|\cdot\|$ attains a minimum and a maximum on S^{n-1} by compactness. Thus there exist positive constants c_1, c_2 such that

$$c_1 \le ||x|| \le c_2$$

for all $x \in S^{n-1}$.

Now for any $x \in \mathbb{R}^n \setminus \{0\}$, dividing by $||x||_2$ gives a point that lies on S^{n-1} . Thus

$$c_1 \le \left\| \frac{x}{\|x\|_2} \right\| \le c_2.$$

By homogeneity (N3),

$$c_1 ||x||_2 \le ||x|| \le c_2 ||x||_2.$$

This is also trivially true for x = 0.

Thus
$$\|\cdot\| \sim \|\cdot\|_2$$
.

Remark. The idea of the proof is as follows.

Any homogenous function is determined by its value on the unit sphere. A homogenous function of degree zero is essentially nothing but a function on the unit sphere $(f(v) = f(\widehat{v}))$.

The function $x \mapsto \frac{\|x\|}{\|x\|_2}$ is a continuous homogenous function on degree 0. The unit sphere is known to be compact under the Euclidean norm (and every other, but not before we complete the proof). Thus

$$c_1 \le \frac{\|x\|}{\|x\|_2} \le c_2$$

Lecture 2: One norm to rule them all

for some positive constants c_1, c_2 .

Definiteness and \triangle are required for the ratio to be continuous. Homogeneity is required for it to be homogeneous. Is positivity required?

Remark. We technically only need to show $c_1||x||_2 \le ||x||$, since the other inequality is proven in the previous proof. It is nonetheless clearer to show both inequalities.

Exercise 1.17 (Self). Show that (N1) follows from (N3) and (N4).

Solution. Let $v \in V$. By triangle inequality, $||v|| = ||-v + 2v|| \le ||-v|| + ||2v||$. By homogeneity, this is 3||v||. Thus $||v|| \le 3||v||$, so $||v|| \ge 0$.

Remarks (Finite-dimensional vector spaces).

- Let V be a vector space over \mathbb{R} with dimension $n < \infty$. Using a basis for V, any norm on V induces a norm on \mathbb{R}^n , and vice versa. Norms on V are in a one-to-one correspondence with norms on \mathbb{R}^n .
- Thus any two norms on V are equivalent.
- Any two inner products on V will also be equivalent due to this.
- Any finite-dimensional vector space over \mathbb{R} is complete.

Exercise 1.18. Let $f: \mathbb{R} \setminus \{0\} \to \mathbb{R}$ be given by $f(x) = \frac{1}{x}$. Show that f is continuous. What is the key idea of your proof?

Solution. Let $x_0 \in \mathbb{R} \setminus \{0\}$ and $\varepsilon > 0$. Choose $\delta = \min\{\varepsilon \cdot \frac{1}{2}|x_0|^2, \frac{1}{2}|x_0|\}$. Then for any x in the δ -neighbourhood of x_0 ,

$$|f(x) - f(x_0)| = \left| \frac{1}{x} - \frac{1}{x_0} \right|$$

$$= \frac{|x_0 - x|}{|x||x_0|}$$

$$< \frac{\delta}{|x||x_0|}$$

$$< \frac{2\delta}{|x_0|^2}$$

$$< \varepsilon$$

Remark. The proof works by bounding $\frac{1}{|x|}$. The rest goes to zero as $x \to a$. We will do a similar proof in proposition 1.39.

On \mathbb{R}^n , we will always fix the ℓ^2 -norm

Lecture 3. Wednesday August 7

Notation.

$$L(\mathbb{R}^n, \mathbb{R}^m) = \{T \colon \mathbb{R}^n \to \mathbb{R}^m \mid T \text{ is linear}\}$$

and

$$M_{m \times n}(\mathbb{R}) \cong L(\mathbb{R}^n, \mathbb{R}^m)$$

using the isomorphism $A \mapsto T_A$ where

$$T_A \colon \mathbb{R}^n \to \mathbb{R}^m$$

 $v \mapsto Av$,

where v is interpreted as a column vector. We will also write $L(\mathbb{R}^n)$ for $L(\mathbb{R}^n, \mathbb{R}^n)$.

Definition 1.19 (liminf and limsup). Let $f: X \to \mathbb{R}$ be a function on a topological space X. We define the *limit inferior* and *limit superior* of f as

$$\liminf_{x \to a} f(x) = \sup_{V} \inf_{x \in V} f(x)$$
$$\limsup_{x \to a} f(x) = \inf_{V} \sup_{x \in V} f(x)$$

where V ranges over all open neighbourhoods of a that contain at least one point other than a.

Exercise 1.20 (self). Let (X, d) be a connected metric space with at least two points. Then

$$\liminf_{x \to a} f(x) = \lim_{\varepsilon \searrow 0} \inf_{0 < d(x,a) < \varepsilon} f(x)$$

$$\limsup_{x \to a} f(x) = \lim_{\varepsilon \searrow 0} \sup_{0 < d(x,a) < \varepsilon} f(x)$$

Proof. We first need to show that $\{x \in X \mid 0 < d(x,a) < \varepsilon\}$ is non-empty for each $\varepsilon > 0$. Suppose this were not the case for some ε . Then $B(a,\varepsilon) = \{a\} = \overline{B(a,\varepsilon/2)}$ is clopen, contradicting the connectedness of X. Notice that for any $\varepsilon_1 > \varepsilon_2 > 0$,

$${x \in X \mid 0 < d(x, a) < \varepsilon_1} \subseteq {x \in X \mid 0 < d(x, a) < \varepsilon_2},$$

so the infimum increases as $\varepsilon \searrow 0$.

Definition 1.21 (O notation). Let $f: \mathbb{R}^n \to \mathbb{R}^m$ and $g: \mathbb{R}^n \to \mathbb{R}^k$. We say that

(i) f(x) = o(g(x)) as $x \to a$ if $\lim_{x \to a} \frac{1}{x}$

$$\lim_{x \to a} \frac{\|f(x)\|}{\|g(x)\|} = 0,$$

(ii) f(x) = O(g(x)) as $x \to a$ if

$$\limsup_{x \to a} \frac{\|f(x)\|}{\|g(x)\|} < \infty.$$

where the assumption is that g is non-zero in some neighbourhood of a.

Exercise 1.22. Show that the definition of O is equivalent to the following: We say that f(x) = O(g(x)) as $x \to a$ if there exists an open neighbourhood V of a such that $\frac{\|f(x)\|}{\|g(x)\|}$ is bounded on V.

Solution. Call the ratio h.

$$\begin{split} \inf_{V} \sup_{x \in V} h(x) & \leq \infty \iff \exists V (\sup_{x \in V} h(x) < \infty) \\ & \iff \exists V \exists M (\forall x \in V, h(x) \leq M) \end{split}$$

Exercise 1.23.

1.1 Matrix norms

Definition 1.24 (Hilbert-Schmidt norm). For a matrix $A \in M_{m \times n}(\mathbb{R})$, we define the *Hilbert-Schmidt* or *Frobenius* norm by

$$||A||_{HS} = \left(\sum_{i,j} a_{ij}^2\right)^{1/2}$$

Exercise 1.25. Show that $||A||_{HS}^2 = \operatorname{Tr}(A^{\top}A) = \operatorname{Tr}(AA^{\top})$.

Solution.

$$(A^{\top}A)_{ii} = \sum_{k} (A^{\top})_{ik} A_{ki}$$
$$= \sum_{k} a_{ki}^{2}$$
$$\Longrightarrow \operatorname{Tr}(A^{\top}A) = \sum_{i} \sum_{k} a_{ki}^{2}$$
$$= ||A||_{HS}^{2}.$$

Lecture 3: Big-Oh and matrix norms

Since $||A||_{HS} = ||A^{\top}||_{HS}$, we also have $\text{Tr}(AA^{\top}) = ||A||_{HS}^{2}$.

Proposition 1.26. Any linear transformation is continuous.

Proof. Let $T \in L(\mathbb{R}^n, \mathbb{R}^m)$. Then

$$||Tx|| = ||T(\sum x_i e_i)||$$

$$= ||\sum x_i T e_i||$$

$$\leq \sum |x_i| ||T e_i||$$

$$\leq ||x|| \sum ||T e_i||$$

$$= M||x||$$
(1.2)

where $M = ||Te_1|| + \cdots + ||Te_n||$. Now

$$||Tx - Ty|| = ||T(x - y)|| \le M||x - y||$$

says that T is Lipschitz continuous with Lipschitz constant M.

We temporarity define two norms on $M_{m\times n}(\mathbb{R})$:

$$\begin{split} \|T\|_S &= \sup_{\|x\|=1} \|Tx\| \\ \|T\|_B &= \sup_{\|x\| \le 1} \|Tx\| \end{split}$$

Lemma 1.27. $||T||_S = ||T||_B$.

Proof. From the definition it is obvious that $||T||_S \leq ||T||_B$. Now for any $x \in \mathbb{R}^n \setminus \{0\}$, let y = x/||x||.

$$\begin{split} \|Ty\| &\leq \|T\|_S \\ \frac{\|Tx\|}{\|x\|} &\leq \|T\|_S \\ \Longrightarrow \|Tx\| &\leq \|T\|_S \|x\| \end{split}$$

Thus for $\|x\| \le 1$, we have $\|Tx\| \le \|T\|_S$ (check 0 separately). So $\|T\|_B \le \|T\|_S$.

Definition 1.28 (Operator norm). For any $T \in L(\mathbb{R}^n, \mathbb{R}^m)$, we define the *operator norm* by

$$||T|| = \sup_{||x||=1} ||Tx||$$

From the previous lemma, we can also write

$$||T|| = \sup_{||x|| \le 1} ||Tx|| = \sup_{x \ne 0} \frac{||Tx||}{||x||}.$$

Lecture 3: Big-Oh and matrix norms

From equation (1.2), we have

$$||T|| < ||Te_1|| + \cdots + ||Te_n||.$$

So the operator norm is finite.

Proposition 1.29. The operator norm is a norm on $L(\mathbb{R}^n, \mathbb{R}^m)$.

Proof. Let $T, S \in L(\mathbb{R}^n, \mathbb{R}^m)$.

- (N1) Positivity is by positivity of the vector norm.
- (N2) Suppose T is not identically zero. Let $v \neq 0$ be such that $||Tv|| \neq 0$. Then

$$||T|| = \sup_{x \neq 0} \frac{||Tx||}{||x||} \ge \frac{||Tv||}{||v||} > 0.$$

(N3) $\|\lambda T\| = \sup_{\|x\|=1} \|\lambda Tx\| = |\lambda| \sup_{\|x\|=1} \|Tx\| = |\lambda| \|T\|.$

(N4)

$$||T + S|| = \sup_{\|x\|=1} ||(T + S)x||$$

$$\leq \sup_{\|x\|=1} ||Tx|| + ||Sx||$$

$$\leq \sup_{\|x\|=1} ||Tx|| + \sup_{\|x\|=1} ||Sx||$$

$$= ||T|| + ||S||.$$

Proposition 1.30. Let $T_2 \in L(\mathbb{R}^m, \mathbb{R}^n)$ and $T_1 \in L(\mathbb{R}^n, \mathbb{R}^k)$. Then

$$||T_1 \circ T_2|| \le ||T_1|| ||T_2||$$

Proof. Let $x \in \mathbb{R}^m$ with ||x|| = 1. Then

$$||T_1T_2x|| \le ||T_1|| ||T_2x|| \le ||T_1|| ||T_2||.$$

Since $M_{m\times n}(\mathbb{R}) \cong \mathbb{R}^{mn}$, we can conclude that the Hilbert-Schmidt norm and the operator norm are equivalent, as are any two norms on $M_{m\times n}(\mathbb{R})$. Thus we can talk about openness and continuity without specifying the norm. Problem 1.10 discusses their equivalence with specific bounds.

Proposition 1.31. $GL_n(\mathbb{R})$ is open in $M_n(\mathbb{R})$.

Proof. det: $M_n(\mathbb{R}) \to \mathbb{R}$ is continuous because it is a polynomial in the entries of the matrix. Note that $GL_n(\mathbb{R}) = \det^{-1}(\mathbb{R} \setminus \{0\})$, so it is the preimage of an open set, which is open by proposition 1.37.

Determinants pose a problem in infinite dimensions. This also doesn't provide estimates on the size of the neighbourhood. We will go through Rudin's proof in proposition 1.39 which avoids determinants.

Meanwhile, here is a special case.

Lemma 1.32. There is an open ball around I in $M_n(\mathbb{R})$ that is contained in $GL_n(\mathbb{R})$.

Proof. A reasonable guess for the radius is 1 (intuiting from the 1D case). Let $X \in M_n(\mathbb{R})$ with ||X - I|| < 1.

Let $v \in \mathbb{R}^n \setminus \{0\}$. Then ||(X - I)v|| < ||v|| implies that $(X - I)v \neq v$. Thus $Xv \neq 0$ and so X is invertible.

This will be useful in proposition 1.39. This can also be proven by borrowing the following result from \mathbb{C} .

$$\frac{1}{1-z} = 1 + z + z^2 + \dots$$
 for $|z| < 1$.

(This was the first thought a student had when prompted.) We approach it this way as well, since this gives us an explicit inverse.

Lemma 1.33. Let $Z \in M_n(\mathbb{R})$ be such that ||Z|| < 1. Then

- (i) $\sum_{n=0}^{\infty} Z^n$ converges.
- (ii) I Z is invertible.
- (iii) $(I Z)^{-1} = \sum_{n=0}^{\infty} Z^n$.

Proof. By $||AB|| \le ||A|| ||B||, ||Z^k|| \le ||Z||^k$.

It is easy to see that the series converges by the Cauchy criterion. For any $\varepsilon > 0$, there is some n such that

$$\left\| \sum_{k=n}^{m} Z^k \right\| \le \sum_{k=n}^{m} \|Z^k\| < \varepsilon$$

for all m > n.

To see that I - Z is invertible, note that

$$||I - Z|| \ge ||I|| - ||Z||| = 1 - ||Z|| > 0.$$

Finally, let $S_n = \sum_{k=0}^n Z^k$ and $S_\infty = \lim_{n \to \infty} S_n$. Then $(I-Z)S_n = I-Z^{n+1}$ and so $(I-Z)S_n \to I$ as $n \to \infty$. Since matrix multiplication is continuous, we can take the limit inside the product and get $(I-Z)S_\infty = I$.

Remark. For infinite-dimensional spaces, we also need to show $S_{\infty}(I-Z) = I$, which will be done in the exact same way.

Proposition 1.34. $A \mapsto A^{-1}$ is continuous on $GL_n(\mathbb{R})$.

Proof. Let $A \in GL_n(\mathbb{R})$. Then $A^{-1} = \frac{1}{\det A} \operatorname{adj} A$. Each entry of A^{-1} is a rational function in the entries of A, so $A \mapsto A^{-1}$ is continuous by exercise 1.35.

Exercise 1.35. Let $U \subseteq \mathbb{R}^n$ be an open set. Let $f: U \to \mathbb{R}^m$ be such that

$$f(x) := (f_1(x), f_2(x), \dots, f_n(x)), \quad x \in U$$

Lecture 4. Friday August 9

Show that f is continuous at $a \in U$ iff each f_i is continuous at a.

Solution. Consider the ℓ^1 norm on \mathbb{R}^m .

Suppose f is continuous. Since $|f_1(x) - f_1(y)| \le ||f(x) - f(y)||$, so is each f_i .

Suppose each f_i is continuous at a. For any $\varepsilon > 0$, there exists $\delta_i > 0$ such that $|f_i(x) - f_i(a)| < \frac{1}{m}\varepsilon$ in a δ_i -neighbourhood of a. Let $\delta = \min\{\delta_1, \delta_2, \ldots, \delta_n\}$.

Exercise 1.36. Let f(x) = o(g(x)) and g(x) = O(h(x)). Then show that f(x) = o(h(x)).

Solution.

$$\limsup_{x \to a} \frac{\|f(x)\|}{\|h(x)\|} < c \quad \text{and} \quad \lim_{x \to a} \frac{\|g(x)\|}{\|h(x)\|} = 0$$

Thus

$$\limsup_{x \to a} \frac{\|f(x)\|}{\|h(x)\|} = 0.$$

Proposition 1.37. Suppose X and Y are metric spaces. Then the following are equivalent.

- (i) f is continuous.
- (ii) $f^{-1}(V)$ is open whenever V is open in Y.

Solution. Suppose f is continuous. Let $V \subseteq Y$ be open.

Let $x \in f^{-1}(V)$. Then $f(x) \in V$. There is some $\varepsilon > 0$ such that $B(f(x), \varepsilon) \subseteq V$. But by continuity, there is some $\delta > 0$ such that $f(B(x, \delta)) \subseteq B(f(x), \varepsilon) \subseteq V$. Thus $B(x, \delta) \subseteq f^{-1}(V)$.

Conversely, suppose $f^{-1}(V)$ is open whenever V is open. Then for any $x \in X$ and $\varepsilon > 0$, we have that $f^{-1}(B(f(x), \varepsilon))$ is open. So some δ -neighbourhood of x in X that is contained in $f^{-1}(B(f(x), \varepsilon))$.

Assignment 1

Problem 1.1. Let $(V, \|\cdot\|)$ be a normed linear space.

- up August 2 due August 12 quiz August 14
- (i) Show that the addition map $(u, v) \mapsto u + v$ is continuous.
- (ii) Show that the scalar multiplication map $(\alpha, u) \mapsto \alpha u$ is continuous.

Proof.

(i)
$$||u_2 + v_2 - (u_1 + v_1)|| \le ||u_2 - u_1|| + ||v_2 - v_1||$$
.

(ii)
$$\|\alpha_2 u_2 - \alpha_1 u_1\| = \|\alpha_2 u_2 - \alpha_1 u_2 + \alpha_1 u_2 - \alpha_1 u_1\| = \|(\alpha_2 - \alpha_1)u_2 + \alpha_1(u_2 - u_1)\| \le |\alpha_2 - \alpha_1| \|u_2\| + |\alpha_1| \|u_2 - u_1\|.$$

Problem 1.2. Let $(V, \|\cdot\|)$ be a normed linear space. Prove that

$$|||x|| - ||y||| \le ||x - y||$$

for all $x, y \in V$. Show that the function $x \mapsto ||x||$ from V to \mathbb{R} is continuous.

Proof. By the \triangle inequality,

$$||x|| = ||x - y + y|| \le ||x - y|| + ||y|| \implies ||x|| - ||y|| \le ||x - y||.$$

Similarly

$$||y|| = ||y - x + x|| \le ||y - x|| + ||x|| \implies ||x|| - ||y|| \ge -||x - y||.$$

Thus

$$|||x|| - ||y||| \le ||x - y||$$

 $\Big|\|x\|-\|y\|\Big|\leq \|x-y\|.$ To show that $\|\cdot\|$ is continuous, do what exactly? Notice

$$|||x|| - ||y||| \le ||x - y||?$$

Problem 1.3. For $x, y \in \mathbb{R}^n$, show that

$$|\langle x, y \rangle| \le ||x||_2 ||y||_2 \tag{1.3}$$

Also show that the two sides in equation (1.3) are equal if and only if x and y are linearly dependent over \mathbb{R} .

Proof. If either of x or y is 0, both sides are 0. Suppose $x, y \neq 0$. Let $\hat{x} = \frac{x}{\|x\|_2}$ and $\hat{y} = \frac{y}{\|y\|_2}$. Then proving equation (1.3) amounts to proving

$$|\langle \hat{x}, \hat{y} \rangle| \leq 1$$

because of homogeneity of the inner product.

$$0 \leq \sum_{i=1}^{n} (\widehat{x}_i - \widehat{y}_i)^2$$

$$0 \leq \sum_{i=1}^{n} \widehat{x}_i^2 - 2\widehat{x}_i \widehat{y}_i + \widehat{y}_i^2$$

$$2 \sum_{i=1}^{n} \widehat{x}_i \widehat{y}_i \leq \sum_{i=1}^{n} \widehat{x}_i^2 + \sum_{i=1}^{n} \widehat{y}_i^2$$

$$\langle \widehat{x}, \widehat{y} \rangle \leq 1.$$

Similarly $\langle -\hat{x}, \hat{y} \rangle \leq 1$, which gives $\langle \hat{x}, \hat{y} \rangle \geq -1$.

Problem 1.4. Let $\{x_k\}_{k\in\mathbb{N}}\subseteq\mathbb{R}^n$ and $x\in\mathbb{R}^n$. Show that $\{x_k\}_{k\in\mathbb{N}}$ converges to x if and only if $\{\langle x_k, y\rangle\}$ converges to $\langle x, y\rangle$ for all $y\in\mathbb{R}^n$.

Proof. Suppose $x_k \to x$. Let $y \in \mathbb{R}^n$. Then

$$|\langle x_k, y \rangle - \langle x, y \rangle| = |\langle x_k - x, y \rangle| \le ||x_k - x|| ||y|| \to 0.$$

Conversely, suppose $\langle x_k, y \rangle \to \langle x, y \rangle$ for all $y \in \mathbb{R}^n$. Then $\langle x_k, e_i \rangle \to \langle x, e_i \rangle$ for all i. Thus $x_k \to x$ componentwise.

Problem 1.5. Let $1 < p, q < \infty$ be such that $\frac{1}{p} + \frac{1}{q} = 1$. Show that for any $a \ge 0$ and $x \ge 0$ the following holds:

$$xa \le \frac{a^p}{p} + \frac{x^q}{q}.\tag{1.4}$$

Show that in equation (1.4) equality holds if and only if $x^q = a^p$.

Proof. Let $a \ge 0$ be fixed. Define $f(x) = xa - \frac{a^p}{p} - \frac{x^q}{q}$. This is differentiable on $[0, \infty)$ since q > 0. $f'(x) = a - x^{q-1}$. Thus

$$f'(x) \le 0 \iff x^{q-1} \le a$$

 $\iff x^{q/p} \le a$
 $\iff x^q \le a^p.$

Thus f is decreasing on $[a^{p/q}, \infty)$ and increasing on $[0, a^{p/q}]$. Thus $f(x) \ge f(a^{p/q}) = 0$. Moreover, since $f'(x) \ne 0$ for $x^q \ne a^p$, we have $f(x) = 0 \iff x^q = a^p$.

Thus
$$xa \leq \frac{a^p}{p} + \frac{x^q}{q}$$
 with equality only if $x^q = a^p$.

Problem 1.6. For $1 \le p \le \infty$ and $x = (x_1, x_2, \dots, x_n)$, we define

$$||x||_{p} = \begin{cases} \left(|x_{1}|^{p} + |x_{2}|^{p} + \dots + |x_{n}|^{p}\right)^{\frac{1}{p}} & 1 \leq p < \infty \\ \max_{1 \leq i \leq n} |x_{i}| & p = \infty \end{cases}$$

(i) Let
$$1 \le q \le \infty$$
 be such that $\frac{1}{p} + \frac{1}{q} = 1$. For any $x, y \in \mathbb{R}^n$, show that $|\langle x, y \rangle| \le ||x||_p ||y||_q$ and $||x + y||_p \le ||x||_p + ||y||_p$. (1.5)

- (ii) Show that $\|\cdot\|_p$ defines a norm on \mathbb{R}^n .
- (iii) Show that $||x||_{\infty} = \lim_{p \to \infty} ||x||_p$ for any $x \in \mathbb{R}^n$.

Proof. We first deal with the case $p = \infty$ for parts (a) and (b).

(i) q = 1.

$$\begin{aligned} |\langle x, y \rangle| &= |x_1 y_1 + x_2 y_2 + \dots + x_n y_n| \\ &\leq |x_1| |y_1| + |x_2| |y_2| + \dots + |x_n| |y_n| \\ &= \max_{1 \leq i \leq n} |x_i| (|y_1| + |y_2| + \dots + |y_n|) \\ &= ||x||_{\infty} ||y||_1 \end{aligned}$$

and

$$||x + y||_{\infty} = \max_{1 \le i \le n} |x_i + y_i|$$

$$\leq \max_{1 \le i \le n} (|x_i| + |y_i|)$$

$$\leq \max_{1 \le i, j \le n} (|x_i| + |y_j|)$$

$$= \max_{1 \le i \le n} |x_i| + \max_{1 \le j \le n} |y_j|$$

$$= ||x||_{\infty} + ||y||_{\infty}.$$

(ii) We have positivity by definition. $||x||_p = 0 \iff \max_{1 \le i \le n} |x_i| = 0 \iff |x_1| = |x_2| = \cdots = |x_n| = 0 \iff x = 0$, so definiteness holds. Homogeneity is since

$$\|\alpha x\|_{\infty} = \max_{1 \le i \le n} |\alpha x_i| = |\alpha| \max_{1 \le i \le n} |x_i| = |\alpha| \|x\|_{\infty}.$$

Triangle inequality is proven above.

Thus $\|\cdot\|_{\infty}$ is a norm.

Now we deal with the case $1 \le p < \infty$.

(i) For $|\langle x, y \rangle| \leq ||x||_p ||y||_q$, we only concern ourselves with $1 < p, q < \infty$. The case p = 1 requires $q = \infty$, which is covered above with p and q interchanged. We will show that the ratio of the two sides is bounded

by 1.

$$\frac{|\langle x, y \rangle|}{\|x\|_p \|y\|_q} = \left| \frac{x_1 y_1 + x_2 y_2 + \dots + x_n y_n}{\|x\|_p \|y\|_q} \right| \\
\leq \sum_{i=1}^n \frac{|x_i| |y_i|}{\|x\|_p \|y\|_q} \\
\leq \sum_{i=1}^n \left(\frac{1}{p} \frac{|x_i|^p}{\|x\|_p^p} + \frac{1}{q} \frac{|y_i|^q}{\|y\|_q^q} \right) \qquad \text{(by equation (1.4))} \\
= \frac{1}{p} \frac{\sum_i |x_i|^p}{\|x\|_p^p} + \frac{1}{q} \frac{\sum_i |y_i|^q}{\|y\|_q^q} \\
= \frac{1}{p} + \frac{1}{q} \\
= 1.$$

We use this result to prove the triangle inequality. (We did this in a UM 204 assignment last semester, with ample of hints and time to spare.)

$$||x+y||_p^p = \sum_{i=1}^n |x_i + y_i|^p$$

$$= \sum_{i=1}^n |x_i + y_i| |x_i + y_i|^{p-1}$$

$$\leq \sum_{i=1}^n |x_i| |x_i + y_i|^{p-1} + \sum_{i=1}^n |y_i| |x_i + y_i|^{p-1}$$

Let $X = (|x_1|, |x_2|, \dots, |x_n|)$ and $Z = (|x_1+y_1|^{p-1}, |x_2+y_2|^{p-1}, \dots, |x_n+y_n|^{p-1})$. Then by equation (1.4),

$$\sum_{i=1}^{n} |x_i| |x_i + y_i|^{p-1} = |\langle X, Z \rangle|$$

$$\leq ||X||_p ||Z||_q$$

where $q = \frac{p}{p-1}$

$$\leq ||x||_p (|x_1 + y_1|^p + \dots + |x_n + y_n|^p)^{\frac{p}{p-1}}$$

= $||x||_p ||x + y||_p^{p-1}$.

Similarly,

$$\sum_{i=1}^{n} |y_i| |x_i + y_i|^{p-1} \le ||y||_p ||x + y||_p^{p-1}.$$

This gives

$$||x + y||_p^p \le (||x||_p + ||y||_p)||x + y||_p^{p-1}$$
$$||x + y||_p \le ||x||_p + ||y||_p.$$

(ii) Positivity is again by definition. $||x||_p = 0 \iff |x_i|^p = 0$ for all i, which is iff x = 0. Homogeneity is trivial to check.

$$\|\alpha x\|_{p} = (|\alpha x_{1}|^{p} + |\alpha x_{2}|^{p} + \dots + |\alpha x_{n}|^{p})^{\frac{1}{p}}$$

$$= (|\alpha|^{p}|x_{1}|^{p} + |\alpha|^{p}|x_{2}|^{p} + \dots + |\alpha|^{p}|x_{n}|^{p})^{\frac{1}{p}}$$

$$= |\alpha|\|x\|_{p}.$$

Triangle inequality is proven above.

Thus $\|\cdot\|_p$ is a norm.

We now prove part (c). The case x=0 is trivial since $\|x\|_p = \|x\|_\infty = 0$ for any p.

WLOG let $||x||_{\infty} = |x_1| > 0$. Then for $1 \le p < \infty$,

$$||x||_{p} = |x_{1}| \left(1 + \frac{|x_{2}|^{p}}{|x_{1}|^{p}} + \dots + \frac{|x_{n}|^{p}}{|x_{1}|^{p}} \right)^{\frac{1}{p}}$$

$$\leq |x_{1}| \cdot n^{\frac{1}{p}}$$

Further,

$$||x||_p = (|x_1|^p + |x_2|^p + \dots + |x_n|^p)^{\frac{1}{p}} \ge (|x_1|^p)^{\frac{1}{p}} = |x_1|.$$

Thus

$$|x_1| \le ||x||_p \le n^{\frac{1}{p}} |x_1|.$$

As $p \to \infty$, $n^{\frac{1}{p}} \to 1$. Thus by the squeeze theorem, $||x||_p \to |x_1| = ||x||_{\infty}$.

Problem 1.7. Let C[a, b] be the set of all complex-valued continuous functions on [a, b].

- (i) Let $f \in C[a,b]$ be such that f is non-negative and $\int_a^b f(x) dx = 0$. Show that $f \equiv 0$.
- (ii) For $f \in C[a,b]$, define

$$||f||_{\infty} := \sup_{x \in [a,b]} |f(x)|, \qquad ||f||_{1} := \int_{a}^{b} |f(x)| \, \mathrm{d}x.$$

Show that $\|\cdot\|_{\infty}$ and $\|\cdot\|_{1}$ are norms on C[a,b].

(iii) Are the above two norms on C[a,b] equivalent? Are they comparable? Solution.

(i) Suppose f is non-zero at some point $c \in [a, b]$. By continuity, $f(x) \ge \frac{f(c)}{2}$ in some neighbourhood $[c - \delta, c + \delta]$. Then f is lower bounded by the step function

$$g(x) = \begin{cases} \frac{f(c)}{2} & x \in [c - \delta, c + \delta] \\ 0 & \text{otherwise} \end{cases}$$

which has positive integral. This would force $\int_a^b f(x) dx > 0$. Contradiction! Such a c cannot exist.

- (ii) Clearly both are non-negative. $||f||_{\infty} = 0 \iff |f(x)| \le 0$ for all $x \in [a, b]$, which is iff $f \equiv 0$. Definiteness of $||\cdot||_1$ is by the previous part. Homogeneity is obvious. Triangle inequality is an extension of the triangle inequality for complex numbers.
- (iii) They are *not* equivalent. Consider [a,b]=[0,1] and $f(x)=e^{-\lambda x}$. Then $\|f\|_{\infty}=1$ and $\|f\|_{1}=\frac{1-e^{-\lambda}}{\lambda}$. One can choose λ to make $\|f\|_{1}$ arbitrarily close to 0. Thus there are no constants $c_{1},c_{2}>0$ such that

$$c_1 ||f||_{\infty} \le ||f||_1 \le c_2 ||f||_{\infty}.$$

However, we can compare the norms as

$$||f||_1 \leq (b-a)||f||_{\infty}.$$

This is simply by noticing that the constant function $x \mapsto \|f\|_{\infty}$ upper bounds |f(x)| and has integral $(b-a)\|f\|_{\infty}$ over [a,b].

Problem 1.8. For $A \in L(\mathbb{R}^n, \mathbb{R}^m)$, let ||A|| denote the operator norm of A. Show that

$$||A|| = \inf\{M : ||Ax|| \le M||x|| \text{ for all } x \in \mathbb{R}^n\}.$$

Proof. $||Ax|| \le M||x||$ is trivially true for x = 0 no matter what M is. Thus

$$\inf\{M : \|Ax\| \le M \|x\| \text{ for all } x \in \mathbb{R}^n\}$$

$$= \inf\{M : \|Ax\| \le M \|x\| \text{ for all } x \in \mathbb{R}^n \setminus \{0\}\}$$

$$= \inf\{M : \left\|A\frac{x}{\|x\|}\right\| \le M \text{ for all } x \in \mathbb{R}^n \setminus \{0\}\}$$

$$= \inf\{M : \|Ay\| \le M \text{ for all } y \in S^{n-1}\}$$

$$= \inf\{\text{upper bounds of } \{\|Ay\| : y \in S^{n-1}\}\}$$

$$= \sup\{\|Ay\| : y \in S^{n-1}\}$$

$$= \|A\|.$$

Problem 1.9. Let A be a real symmetric $n \times n$ matrix.

(i) Show that all eigenvalues of A are real.

(ii) For $1 \le i \le n$, let λ_i denote the eigenvalues of A. Show that

$$||A|| = \max_{1 \le i \le n} |\lambda_i|.$$

Solution.

(i) View A as a linear operator on \mathbb{C}^n . Let λ be an eigenvalue of A and v be the corresponding eigenvector. Then

$$\lambda \langle v, v \rangle = \langle Av, v \rangle = \langle v, Av \rangle = \overline{\lambda} \langle v, v \rangle.$$

Thus $\lambda = \bar{\lambda}$ is real.

(ii) (assuming spectral theorem) WLOG let $\lambda_1 = \max_{1 \leq i \leq n} |\lambda_i|$. Write any vector $x \in \mathbb{R}^n$ as a linear combination of orthonormal eigenvectors $x = \sum_{i=1}^n c_i v_i$, where v_i is the eigenvector corresponding to λ_i . Then $Ax = \sum_{i=1}^n c_i \lambda_i v_i$.

$$||Ax||^{2} = \sum_{i=1}^{n} c_{i}^{2} \lambda_{i}^{2}$$

$$\leq \lambda_{1}^{2} \sum_{i=1}^{n} c_{i}^{2}$$

$$= \lambda_{1}^{2} ||x||^{2}.$$

Thus $||A|| \le \lambda_1$. Moreover, $||Av_1|| = |\lambda_1| ||v_1||$. Thus $||A|| \ge \lambda_1$.

Problem 1.10. Let $A \in L(\mathbb{R}^n, \mathbb{R}^m)$ and $B \in L(\mathbb{R}^k, \mathbb{R}^n)$. Show that

$$||A|| \le ||A||_{HS} \le \sqrt{n}||A||$$
 and $||AB||_{HS} \le ||A||_{HS}||B||_{HS}$.

Proof. $||A||_{HS} = \sqrt{\text{Tr}(A^{\top}A)}$. Recall that the trace of a matrix is the sum of its eigenvalues.

Let v_1, v_2, \ldots, v_n be orthonormal eigenvectors of $A^{\top}A$ with eigenvalues $\lambda_1 \geq \lambda_2 \geq \cdots \geq \lambda_n$ (spectral theorem). Each λ_i is non-negative, since $\langle A^{\top}Ax, x \rangle = \langle Ax, Ax \rangle \geq 0$.

Then for any $x = \sum_{i=1}^{n} c_i v_i$ with ||x|| = 1,

$$||Ax||^2 = \langle Ax, Ax \rangle = \langle A^{\top}Ax, x \rangle = \sum_{i=1}^n c_i^2 \lambda_i \le \lambda_1$$

where the equality holds for $x = v_1$. Thus $||A|| = \sqrt{\lambda_1}$. Since $||A||_{HS}^2 = \sum_{i=1}^n \lambda_i$, we have $\lambda_1 \leq ||A||_{HS}^2 \leq n\lambda_1$. This gives $||A|| \leq ||A||_{HS} \leq \sqrt{n} ||A||$. For $1 \leq i \leq m$ and $1 \leq j \leq k$ let

$$a_i = \begin{pmatrix} A_{i1} & A_{i2} & \cdots & A_{in} \end{pmatrix}^{\mathsf{T}}, \qquad b_j = \begin{pmatrix} B_{1j} \\ B_{2j} \\ \vdots \\ B_{nj} \end{pmatrix}.$$

Assignment 1 — due August 12

Then

$$AB = \begin{pmatrix} \langle a_1, b_1 \rangle & \langle a_1, b_2 \rangle & \cdots & \langle a_1, b_k \rangle \\ \langle a_2, b_1 \rangle & \langle a_2, b_2 \rangle & \cdots & \langle a_2, b_k \rangle \\ \vdots & \vdots & \ddots & \vdots \\ \langle a_m, b_1 \rangle & \langle a_m, b_2 \rangle & \cdots & \langle a_m, b_k \rangle \end{pmatrix}$$

so by Cauchy-Schwarz

$$||AB||_{HS}^{2} = \sum_{i=1}^{m} \sum_{j=1}^{k} \langle a_{i}, b_{j} \rangle^{2}$$

$$\leq \sum_{i=1}^{m} \sum_{j=1}^{k} ||a_{i}||^{2} ||b_{j}||^{2}$$

$$= \left(\sum_{i=1}^{m} ||a_{i}||^{2} \right) \left(\sum_{j=1}^{k} ||b_{j}||^{2} \right)$$

$$= ||A||_{HS}^{2} ||B||_{HS}^{2}.$$

Remark. A far simpler proof that I missed is the following.

$$||Ax||^{2} \leq \sum_{i} \langle a_{i}, x \rangle^{2} \qquad ||A||_{HS}^{2} = \sum_{j} \sum_{i} a_{ij}^{2}$$

$$\leq \sum_{i} ||a_{i}||^{2} ||x||^{2} \qquad = \sum_{j} ||Ae_{j}||^{2}$$

$$= ||A||_{HS}^{2} ||x||^{2} \qquad \leq \sum_{j} ||A||^{2}$$

$$= n||A||^{2}.$$

Quiz

Problem 1.11. Recall the definition of a homogeneous function. Let $f: \mathbb{R}^n \setminus \{0\} \to \mathbb{R}$ be a continuous, non-vanishing homogeneous function of degree k and $\|\cdot\|$ be a fixed norm on \mathbb{R}^n . Show that there exist positive constants $C_1, C_2 > 0$ such that

$$|C_1||x||^k \le |f(x)| \le C_2||x||^k$$

for every $0 \neq x \in \mathbb{R}^n$.

Proof. Choose $C_1 = \min_{\|x\|=1} |f(x)|$ and $C_2 = \max_{\|x\|=1} |f(x)|$. They exist by compactness of the unit sphere, and are positive since f does not vanish. Then for any $x \neq 0$,

$$|f(x)| = ||x||^k \left| f\left(\frac{x}{||x||}\right) \right|$$

is bounded between $C_1||x||^k$ and $C_2||x||^k$.

Problem 1.12. Let V be a vector space over \mathbb{R} . Let d be the discrete metric on V. Is d induced by a norm on V?

Solution. No. Suppose d(x,y) = ||x-y|| for some norm $||\cdot||$, for all $x,y \in V$. Let $x \neq y$. Then d(x,y) = 1 = ||x-y||. But d(2x,2y) = 1 = ||2x-2y|| = 2||x-y|| = 2. Contradiction!

Problem 1.13. For $A \in L(\mathbb{R}^n, \mathbb{R}^m)$, show that $||A|| = ||A^\top||$.

Proof. Notice by Cauchy-Schwarz that for any vector v in a real inner product space,

$$\|v\| = \sup_{\|w\|=1} \langle w, \, v \rangle.$$

(The supremum is achieved at $v/\|v\|$ for $v \neq 0$.) Then

$$||A|| = \sup_{x \in S^{n-1}} ||Ax||$$

$$= \sup_{x \in S^{n-1}} \sup_{y \in S^{m-1}} \langle y, Ax \rangle$$

$$= \sup_{y \in S^{m-1}} \sup_{x \in S^{n-1}} \langle A^{\top}y, x \rangle$$

$$= \sup_{y \in S^{m-1}} ||A^{\top}y||$$

$$= ||A^{\top}||.$$

Problem 1.14. Find maximum of x + 2y + 3z subject to the condition $x^2 + y^2 + z^2 = 1$.

Solution. The function is continuous and the constraint is compact. Thus a maximum exists.

Let r = (x, y, z) and n = (1, 2, 3). As discussed in the previous problem, $\max_{\|r\|=1} \langle n, r \rangle = \|n\|.$

Thus the maximum is $\sqrt{14}$.

Problem 1.15. See problem 1.9.

Exercise 1.38. Let Z be as in $(I-Z)^{-1} = I + Z + O(Z^2)$ and also $(I-Z)^{-1} = {\bf Lecture 5.} \\ I + Z + o(Z^2).$ August 12

The proof of proposition 1.34 is nice and sweet. However, the proof in Rudin generalises better to infinite dimensions. We thus prove it again.

Proposition 1.39.

- (i) Let $A \in M_n(\mathbb{R})$ be such that ||I A|| < 1. Then $A \in GL_n(\mathbb{R})$.
- (ii) Let $A \in GL_n(\mathbb{R})$ be fixed and let $B \in M_n(\mathbb{R})$ be such that

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

Then $B \in \mathrm{GL}_n(\mathbb{R})$.

(iii) $A \mapsto A^{-1}$ is continuous on $GL_n(\mathbb{R})$.

Remark. The second part shows that $GL_n(\mathbb{R})$ is open in $M_n(\mathbb{R})$.

Proof. We proved the first part earlier in lemma 1.32 and again in lemma 1.33 (let Z = I - A, then I - Z = A).

For the second part, let $A \in GL_n(\mathbb{R})$ be fixed and let $||B-A|| < ||A^{-1}||^{-1}$. We can write B - A as $A(A^{-1}B - I)$. Now

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

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

$$< 1$$

Then by the first part, $A^{-1}B \in GL_n(\mathbb{R})$, so that $B \in GL_n(\mathbb{R})$.

For the last part, we want $B^{-1} \to A^{-1}$ as $B \to A$.

$$B^{-1} - A^{-1} = A^{-1}(A - B)B^{-1}$$
(1.6)

We need to bound $||B^{-1}||$. Let W be an open neighbourhood of A of radius $\frac{1}{2}||A^{-1}||^{-1}$. Then $W \subseteq GL_n(\mathbb{R})$.

For any $B \in W$, $||A - B|| ||A^{-1}|| < \frac{1}{2}$ and

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

$$\le ||B^{-1}|| ||A - B|| ||A^{-1}|| \qquad \text{(by equation (1.6))}$$

$$\le \frac{1}{2} ||B^{-1}||.$$

This bounds $||B^{-1}||$ above by $2||A^{-1}||$. Using equation (1.6) again, we have

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

$$\le 2||A^{-1}||^2 \cdot ||A - B||.$$

As
$$B \to A$$
, $B^{-1} \to A^{-1}$.

Lecture 5: Continuity of the inverse; differentiation

Idea. This is similar in spirit to exercise 1.18.

- Equation (1.6) is similar to taking the common denominator in $\frac{1}{x} \frac{1}{a}$.
- The choice of W is similar to choosing $\delta \leq \frac{1}{2}|a|,$ and leads to an identical bound.

Lecture 5: Continuity of the inverse; differentiation

Chapter 2

Differentiation

Definition 2.1. Let $f: \mathbb{R} \to \mathbb{R}$ be a function. We say that f is differentiable at $a \in \mathbb{R}$ if

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

 $\lim_{h\to 0}\frac{f(a+h)-f(a)}{h}$ exists. We denote this limit by f'(a) and call it the *derivative* of f at a.

This doesn't make sense for $f: \mathbb{R}^n \to \mathbb{R}^m$ when n > 2 (for n = 2 we can identify \mathbb{R}^2 with \mathbb{C}).

Theorem 2.2 (Hurwitz' theorem). \mathbb{R}^n is a

We will redefine differentiability for real functions.

Proposition 2.3. Let U be an open subset of \mathbb{R} and $f: U \to \mathbb{R}$. Let $a \in U$. Then f is differentiable at a if and only if there exists a linear $map \ T \in L(\mathbb{R}, \mathbb{R}) \ such \ that$

$$f(a+h) - f(a) = Th + o(h).$$

Proof. Suppose f is differentiable at $a \in U$.

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

We can rewrite this as

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

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

where $T_{\alpha} \in L(\mathbb{R}, \mathbb{R})$ is the linear map $x \mapsto \alpha x$.

Conversely, suppose there exists a linear map T such that f(a+h) –

$$f(a) - Th = o(h)$$
. Then

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

$$\implies \lim_{h \to 0} \left| \frac{f(a+h) - f(a)}{h} - T(1) \right| = 0$$

$$\implies \lim_{h \to 0} \frac{f(a+h) - f(a)}{h} = T(1).$$

Definition 2.4. Let $U \subseteq \mathbb{R}^n$ be an open set containing a. Let $f \colon U \to \mathbb{R}^m$. We say that f is differentiable at a if there exists a linear map $T \in L(\mathbb{R}^n, \mathbb{R}^m)$ such that

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

We say that T is the *derivative* of f at a and write f'(a) = T. If f is differentiable at every point in U, we say that f is differentiable on U. **Lecture 6.** Monday August 19

Writing f'(a) requires the derivative to be unique.

Proposition 2.5. Let $T_1, T_2 \in L(\mathbb{R}^n, \mathbb{R}^m)$ be satisfying the definition of differentiability at a for $f: U \to \mathbb{R}^m$. Then $T_1 = T_2$.

Proof. Let $T = T_1 - T_2$. Then

$$Th = T_1h - T_2h$$

= $(f(a+h) - f(a) - T_2h) - (f(a+h) - f(a) - T_1h)$
= $o(h) - o(h) = o(h)$.

We have $\lim_{h\to 0} \frac{||Th||}{||h||} = 0$. Let $v \in \mathbb{R}^n \setminus \{0\}$. As $t\to 0$, $tv\to 0$. Thus

$$\lim_{t \to 0} \frac{\|T(tv)\|}{\|tv\|} = \lim_{t \to 0} \frac{|t| \|Tv\|}{|t| \|v\|}$$
$$= \frac{\|Tv\|}{\|v\|} = 0.$$

Thus Tv = 0 for all $v \in \mathbb{R}^n$.

Proposition 2.6. Differentiability at a point implies continuity at that point.

Proof. Suppose f is differentiable at a with f'(a) = T. Let

$$q(h) = f(a+h) - f(a) - Th.$$

$$||f(a+h) - f(a)|| = ||f(a+h) - f(a) - Th + Th||$$

$$\leq ||q(h)|| + ||Th||$$

$$\leq \frac{||q(h)||}{||h||} ||h|| + ||T|| ||h||.$$

As $h \to 0$, each term goes to 0.

For *finding* the derivative, it is helpful to do the following:

- Use little-o notation.
- Identify the linear map T.
- Ignore the little-o terms.

If f(a+h) = f(a) + Th + o(h), then f'(a) = T. Examples.

- Let $f: \mathbb{R}^n \to \mathbb{R}^m$ be given by f(x) = c for some constant $c \in \mathbb{R}^m$. For any $a \in \mathbb{R}^n$, we can write f(a+h) = f(a) + 0 + 0. Thus f'(a) = 0.
- Let $f \in L(\mathbb{R}^n, \mathbb{R}^m)$. Then f(a+h) = f(a) + f(h) + 0. Thus f'(a) = f.
- Let $f: \mathbb{R} \to \mathbb{R}$ be given by f(x) = x. This is a special case of the previous example. $f'(a) = \mathrm{id}$. Thus $f'(A)(H) = A^2H + AHA + HA^2$.

Even though we are developing calculus on \mathbb{R}^n , it is trivially extended to all finite-dimensional normed linear spaces over \mathbb{R} via the natural identification with \mathbb{R}^n .

Lecture 7.
Wednesday
August 21

We will continue our examples.

Examples.

• Let $f: \mathbb{R}^2 \to \mathbb{R}$ be given by f(x,y) = xy. Write $a = (a_1, a_2), h = (h_1, h_2)$. Then

$$f(a+h) = f(a_1 + h_1, a_2 + h_2)$$

$$= (a_1 + h_1)(a_2 + h_2)$$

$$= a_1 a_2 + a_1 h_2 + a_2 h_1 + h_1 h_2$$

$$= f(a) + (a_2 \quad a_1) \binom{h_1}{h_2} + o(h).$$

Let us show $h_1h_2 = o(h)$.

$$\frac{|h_1 h_2|}{\|h\|} = |h_1| \frac{|h_2|}{\|h\|} \le |h_1| \to 0.$$

Thus f'(a) is the map $(h_1, h_2) \mapsto a_2h_1 + a_1h_2$. As a matrix, this is $\begin{pmatrix} a_2 & a_1 \end{pmatrix}$.

- Let $f: M_n(\mathbb{R}) \to M_n(\mathbb{R})$ be given by f(X) = X. We could identify $M_n(\mathbb{R})$ with \mathbb{R}^{n^2} and construct a linear map from \mathbb{R}^{n^2} to \mathbb{R}^{n^2} . It is however advisable to construct a linear map from $M_n(\mathbb{R})$ to $M_n(\mathbb{R})$. This is again a specical case of the second example. Thus $f'(A) = f = \mathrm{id}$.
- Let $f: M_n(\mathbb{R}) \to M_n(\mathbb{R})$ be given by $f(X) = X^2$. Then $f(A+H) = (A+H)^2$ $= A^2 + AH + HA + H^2$ = f(A) + AH + HA + o(H)

since

$$\frac{\|H^2\|}{\|H\|} \le \frac{\|H\|^2}{\|H\|} = \|H\| \to 0.$$

Thus f'(A)(H) = AH + HA.

• Let $f: M_n(\mathbb{R}) \to M_n(\mathbb{R})$ be given by $f(X) = X^3$. Then

$$f(A + H) = (A + H)^{3}$$

$$= A^{3} + A^{2}H + AHA + HA^{2}$$

$$+ AH^{2} + HAH + H^{2}A + H^{3}$$

$$= A^{3} + (A^{2}H + AHA + HA^{2}) + o(H).$$

• Let $f: GL_n(\mathbb{R}) \to M_n(\mathbb{R})$ be given by $f(X) = X^{-1}$. Recall that if ||Z|| < 1, then

$$(I-Z)^{-1} = I + Z + O(Z^2) = I + Z + o(Z).$$

Thus for small enough ||H||,

$$(A+H)^{-1} = (A(I+A^{-1}H))^{-1}$$

$$= (I-A^{-1}H+o(-A^{-1}H))A^{-1}$$

$$= A^{-1}-A^{-1}HA^{-1}+o(H).$$
(2.1)

Let us do the o(H) term more carefully. Let $(I + A^{-1}H)^{-1} = I - A^{-1}H + u(H)$ where

$$\lim_{H \to 0} \frac{\|u(H)\|}{\|-A^{-1}H\|} = 0.$$

Then

$$(A+H)^{-1} = (I+A^{-1}H)^{-1}A^{-1}$$

= $(I-A^{-1}H+u(H))A^{-1}$
= $A^{-1}-A^{-1}HA^{-1}+u(H)A^{-1}$.

But

$$\begin{split} \frac{\|u(H)A^{-1}\|}{\|H\|} &\leq \frac{\|u(H)\|}{\|H\|} \|A^{-1}\| \\ &\leq \frac{\|u(H)\|}{\|-A^{-1}H\|} \frac{\|-A^{-1}H\|}{\|H\|} \|A^{-1}\| \\ &\leq \frac{\|u(H)\|}{\|-A^{-1}H\|} \|A^{-1}\|^2 \\ &\to 0. \end{split}$$

Thus from equation (2.1), $f'(A)(H) = -A^{-1}HA^{-1}$.

However, we have can simply use ?? as follows:

$$u(H) = o(-A^{-1}H) = o(O(H)) = o(H).$$

• Let $f: M_n(\mathbb{R}) \to M_n(\mathbb{R})$ be given by $f(X) = X^{-2}$. Then

$$f(A+H) = (A+H)^{-2}$$

= $(A^{-1} - A^{-1}HA^{-1} + o(H))(A^{-1} - A^{-1}HA^{-1} + o(H))$
= $A^{-2} - A^{-2}HA^{-1} - A^{-1}HA^{-2} + o(H)$.

Thus $f'(A)(H) = -A^{-1}(A^{-1}H + HA^{-1})A^{-1}$.

Remarks.

 $\frac{1}{x+h} - \frac{1}{x} = \frac{1}{x+h}(x-(x+h))\frac{1}{x} \to -\frac{1}{x}h\frac{1}{x}$ • 1 1 1 1 (2) (3) (4) (2) 1

 $\frac{1}{(x+h)^2} - \frac{1}{x^2} = \frac{1}{(x+h)^2} (x^2 - (x+h)^2) \frac{1}{x^2}$

Exercise 2.7 (Sum). Let $U \subseteq \mathbb{R}^n$ be open and $a \in U$. Let $f, g: U \to \mathbb{R}^m$ both be differentiable at U. Then f+g is differentiable at a with (f+g)'(a) = August 23 f'(a) + g'(a).

Lemma 2.8. Let $f, g, h : \mathbb{R}^n \to \mathbb{R}^m$. Suppose g = O(f) and h = o(g). Then h = o(f).

Proof.

$$\frac{\|h(x)\|}{\|f(x)\|} = \frac{\|h(x)\|}{\|g(x)\|} \frac{\|g(x)\|}{\|f(x)\|}$$

The first term goes to 0 and the second term is bounded. Thus $\frac{\|h(x)\|}{\|f(x)\|} \to 0$

Proposition 2.9 (chain rule). Let $U \subseteq \mathbb{R}^n$ be open and $a \in U$. Let $f: U \to \mathbb{R}^n$ \mathbb{R}^m be differentiable at a. Let $V \subseteq \mathbb{R}^m$ be an open set containing f(a) and $g: \mathbb{R}^m \to \mathbb{R}^k$ be differentiable at f(a). Then $g \circ f$ is differentiable at a with

$$(g \circ f)'(a) = g'(f(a)) \circ f'(a).$$

Proof. For small enough h,

$$g(f(a+h)) - g(f(a)) = g(f(a) + f'(a)h + u(h)) - g(f(a))$$

= $g'(f(a))[f'(a)h + u(h)] + v(f'(a)h + u(h))$

where $\frac{\|u(h)\|}{\|h\|} \to 0$ and $\frac{\|v(f'(a)h+u(h))\|}{\|f'(a)h+u(h)\|} \to 0$. Call f(a) = b and f'(a)h + u(h) = k(h) for convenience. We have

$$\frac{\|u(h)\|}{\|h\|} \to 0$$
 and $\frac{\|v(k(h))\|}{\|k(h)\|} \to 0$

and

$$q(f(a+h)) - q(f(a)) = q'(b)f'(a)h + q'(b)u(h) + v(k(h)).$$

We need to show

$$g'(b)u(h) + v(k(h)) = o(h).$$

The first term is easy, since $||g'(b)u(h)|| \le ||g'(b)|| ||u(h)||$.

For the second term, we write

$$\frac{\|v(k(h))\|}{\|h\|} = \frac{\|v(k(h))\|}{\|k(h)\|} \frac{\|k(h)\|}{\|h\|}$$

$$\leq \frac{\|v(k(h))\|}{\|k(h)\|} \left(\|f'(a)\| + \frac{\|u(h)\|}{\|h\|} \right)$$

which goes to 0 as the second term is bounded.

Let $\varepsilon > 0$ be given. Then there exists a neighbourhood W_1 of 0 in V on which

$$||v(k)|| \leq \varepsilon ||k||.$$

Since f is continuous, there exists a neighbourhood W_2 of 0 in U such that

$$k(h) = f(a+h) - f(a) \in W_1$$

for each $h \in W_2$. Thus on this neighbourhood W_2 , we have

$$\frac{\|v(k(h))\|}{\|h\|} \le \varepsilon \left(\|f'(a)\| + \frac{\|u(h)\|}{\|h\|} \right).$$

Thus

$$\lim_{h \to 0} \frac{v(k(h))}{h} = 0.$$

How does Rudin deal with the vanishing of k(h)?

Examples.

• Consider $f: GL_n(\mathbb{R}) \to M_n(\mathbb{R})$ and $g: M_n(\mathbb{R}) \to M_n(\mathbb{R})$ given by $f(X) = X^{-1}$ and $g(X) = X^2$.

These are differntiable with $f'(X)(H) = -X^{-1}HX^{-1}$ and g'(X)(H) = XH + HX. By the chain rule, we have

$$(g \circ f)'(X)(H) = g'(f(X))(f'(X)(H))$$

= $X^{-1}(-X^{-1}HX^{-1}) + (-X^{-1}HX^{-1})X^{-1}$
= $-X^{-1}(X^{-1}H + HX^{-1})X^{-1}$.

Definition 2.10 (directional derivative). Let $U \subseteq \mathbb{R}^n$ be open and $a \in U$. Let $f: U \to \mathbb{R}$. For $v \in \mathbb{R}^n$, we define

$$(D_v f)(a) := \lim_{t \to 0} \frac{f(a+tv) - f(a)}{t}$$

to be the directional derivative of f at a in the direction of v.

Exercise 2.11. Show that $D_{\lambda v} = \lambda D_v$.

Solution. If $\lambda = 0$, both sides are 0. Otherwise,

$$(D_{\lambda v}f)(a) = \lim_{t \to 0} \frac{f(a+t\lambda v) - f(a)}{t}$$
$$= \lambda \lim_{t \to 0} \frac{f(a+(\lambda t)v) - f(a)}{\lambda t}$$
$$= \lambda \lim_{t \to 0} \frac{f(a+tv) - f(a)}{t}$$
$$= \lambda (D_v f)(a) = (\lambda D_v f)(a).$$

Definition 2.12 (partial derivative). Let $U \subseteq \mathbb{R}^n$ be open and $a \in U$. Let $f: U \to \mathbb{R}$. We define

$$\frac{\partial f}{\partial x_i}(a) = D_i f(a) = \partial_i f(a) := (D_{e_i} f)(a)$$

to be the partial derivative of f at a with respect to the i-th coordinate.

Observe that if $g = x \mapsto f(a_1, \dots, a_{i-1}, x, a_{i+1}, \dots, a_n)$, then $D_i f(a) = g'(a_i)$.

Lecture 9.

Definition 2.13 (gradient). Let $U \subseteq \mathbb{R}^n$ be open and $f: U \to \mathbb{R}$. We define August 25 the gradient $\nabla f: U \to \mathbb{R}^n$ of f by

$$(\nabla f)(a) := \left(\frac{\partial f}{\partial x_1}(a), \dots, \frac{\partial f}{\partial x_n}(a)\right)$$

for all $a \in U$.

Throughout this lecture, $U \subseteq \mathbb{R}^n$ is open and $a \in U$, and $f: U \to \mathbb{R}^m$.

Proposition 2.14. Suppose $f = (f_1, ..., f_m)$ is differentiable at a. Then $\frac{\partial f_j}{\partial x_i}(a)$ exists for each $i \in [n]$ and $j \in [m]$. Moreover,

$$f'(a)(e_i) = \sum_{j=1}^{m} \frac{\partial f_j}{\partial x_i}(a)e_j$$

for all $i \in [n]$. Equivalently,

$$f'(a) = \begin{pmatrix} \nabla f_1(a) \\ \vdots \\ \nabla f_m(a) \end{pmatrix}.$$

We have abused notation in the above statement to let e_1 be in \mathbb{R}^n or \mathbb{R}^m depending on the context.

Proof. Let $f'(a) = T \in L(\mathbb{R}^n, \mathbb{R}^m)$. For $i \in [n]$,

$$\frac{\partial f_j}{\partial x_i}(a) = \lim_{t \to 0} \frac{f_j(a + te_i) - f_j(a)}{t}$$

$$\implies \frac{\partial f_j}{\partial x_i}(a) - T(e_i)_j = \lim_{t \to 0} \frac{f_j(a + te_i) - f_j(a) - tT(e_i)_j}{t}$$

$$= \lim_{t \to 0} \left(\frac{f(a + te_i) - T(te_i) - f(a)}{t}\right)_j$$

$$= 0.$$

Thus

$$T(e_i) = \sum_{j=1}^{m} \frac{\partial f_j}{\partial x_i}(a)e_j$$

Definition 2.15 (curve). A map $\gamma: (a,b) \to \mathbb{R}^n$ is a *curve* in \mathbb{R}^n . If γ is differentiable, it is a *differentiable curve*.

Remark. Chain rule remains valid for curves. Suppose $f: U \to \mathbb{R}$ and $\gamma: (a,b) \to U$. Then

$$(f \circ \gamma)'(t) = f'(\gamma(t))(\gamma'(t))$$

$$= f'(\gamma(t)) \left(\sum \gamma_i'(t) e_i\right)$$

$$= \sum \gamma_i'(t) f'(\gamma(t))(e_i)$$

$$= \sum \gamma_i'(t) \frac{\partial f}{\partial x_i}(\gamma(t))$$

$$= \nabla f(\gamma(t)) \cdot \gamma'(t).$$

Definition 2.16. Let $\gamma: (-\varepsilon, \varepsilon) \to U$ be such that $\gamma(0) = a$. $\gamma'(0)$ is the tangent vector to γ at a, and $(f \circ \gamma)'(0)$ is the derivative of f along γ at a.

Proposition 2.17. If f is differentiable at a, then for each $v \in \mathbb{R}^n$, $D_v f(a)$ exists and equals $\nabla f(a) \cdot v$.

Proof. Let $\gamma(t) = a + tv$. There exist some $\varepsilon > 0$ such that $\gamma(-\varepsilon, \varepsilon) \subseteq U$ and $\gamma'(0) = v$. Then

$$D_v f(a) = \lim_{t \to 0} \frac{f(a+tv) - f(a)}{t} = (f \circ \gamma)'(0) = \nabla f(a) \cdot v.$$

Examples.

Lecture 10. Wednesday August 28

• Let $f: \mathbb{R}^2 \to \mathbb{R}$ be given by

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

Note that $|f(x,y)| = |x| \frac{y^2}{x^2 + y^2} \le |x|$ for $(x,y) \ne 0$, so f is continuous at (0,0). Suppose f were differentiable at 0. Then the derivative of f at 0 would be

$$\begin{pmatrix} D_1 f(0) & D_2 f(0) \end{pmatrix} = \begin{pmatrix} 0 & 0 \end{pmatrix}.$$

Then

$$0 = \lim_{(h,k)\to 0} \frac{|f(h,k) - f(0,0) - f'(0)(h,k)|}{\|(h,k)\|}$$
$$= \lim_{(h,k)\to 0} \frac{f(h,k)}{\|(h,k)\|}$$
$$= \lim_{(h,k)\to 0} \frac{hk^2}{(h^2 + k^2)^{3/2}}.$$

This is homogenous, and thus constant along lines through the origin.

$$0 = \lim_{t \to 0} \frac{t^3}{(2t)^{3/2}}$$
$$= \frac{1}{2\sqrt{2}}.$$

This is essentially showing that $D_{(1,1)}f(0) \neq f'(0)(1,1)$.

• Let $f: \mathbb{R}^2 \to \mathbb{R}$ be given by

$$f(x,y) = [(x,y) \neq 0](x^2 + y^2) \sin\left(\frac{1}{\sqrt{x^2 + y^2}}\right).$$

f is continuous at 0. We first find the gradient.

$$D_1 f(x, y) = \lim_{x \to 0} \frac{f(x, 0)}{x}$$
$$= \lim_{x \to 0} x \sin\left(\frac{1}{|x|}\right)$$
$$= 0.$$

Similarly,

$$D_2 f(x, y) = 0.$$

Thus if the derivative exists, it must be 0.

$$\lim_{(h,k)\to 0} \frac{|f(h,k) - f(0,0) - f'(0)(h,k)|}{\|(h,k)\|}$$

$$= \lim_{(h,k)\to 0} \|(h,k)\| \sin\left(\frac{1}{\|(h,k)\|}\right)$$

$$= 0.$$

Thus the derivative is indeed 0.

• Let $f: \mathbb{R}^2 \to \mathbb{R}$ be given by

$$f(x,y) = [(x,y) \neq 0]\sqrt{x^2 + y^2} \sin\left(\frac{1}{\sqrt{x^2 + y^2}}\right).$$

Then $f(x,0) = x \sin(\frac{1}{x})$ is not differentiable at 0. Thus f is not differentiable at 0.

Theorem 2.18 (Rolle's theorem). Let $f:[a,b] \to \mathbb{R}$ be continuous on [a,b] and differentiable on (a,b). If f(a)=f(b), then there exists $c \in (a,b)$ such that f'(c)=0.

Theorem 2.19 (mean value theorem). Let $f: [a,b] \to \mathbb{R}$ be continuous on [a,b] and differentiable on (a,b). Then there exists $c \in (a,b)$ such that

$$f(b) - f(a) = f'(c)(b - a).$$

Does an analogue of the mean value theorem hold for several variables? Consider

$$f(t) = (\cos t, \sin t).$$

Then $f(0) = f(2\pi)$, but |f'(t)| = 1 is never 0.

Proposition 2.20. Let $\gamma: [a,b] \to \mathbb{R}^n$ be continuous on [a,b] and differentiable on (a,b). Then there exists $c \in (a,b)$ such that

$$\|\gamma(b) - \gamma(a)\| \le (b-a)\|\gamma'(c)\|.$$

Assignment 2

up August 2 due August 12 quiz August 14