

Analysis II

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1 Vector spaces

1.1 Vector spaces

If $a_n \in \mathbb{R}$, $(a_n) \rightarrow a$ if for every $\epsilon > 0$, $\exists N$ such that $|a_n - a| < \epsilon$ whenever $n > N$.

Now consider a general vector space:

Definition. Let V be a real vector space. A *norm* on V is a function $\|\cdot\| : V \rightarrow \mathbb{R}$ satisfying:

- $\|\mathbf{v}\| \geq 0 \ \forall \mathbf{v} \in V$, and $\|\mathbf{v}\| = 0 \iff \mathbf{v} = \mathbf{0}$;
- $\|\lambda \mathbf{v}\| = |\lambda| \cdot \|\mathbf{v}\|$, $\forall \lambda \in \mathbb{R}$ and $\mathbf{v} \in V$;
- $\|\mathbf{v} + \mathbf{w}\| \leq \|\mathbf{v}\| + \|\mathbf{w}\|$, $\forall \mathbf{v}, \mathbf{w} \in V$ (triangle inequality).

Example. $\|\mathbf{v}\|_2 = (\sum v_i^2)^{\frac{1}{2}}$, the Euclidean norm;

$$\|\mathbf{v}\|_1 = \sum |v_i|;$$

$$\|\mathbf{v}\|_\infty = \max\{|v_1|, \dots, |v_n|\}.$$

Example. Let $V = C[0, 1] = \{f : [0, 1] \rightarrow \mathbb{R} \mid f \text{ is continuous}\}$. Then we can have the following norms:

- $\|f\|_1 = \int_0^1 |f(x)| dx$;
- $\|f\|_2 = \left(\int_0^1 f(x)^2 dx\right)^{\frac{1}{2}}$;
- $\|f\|_\infty = \max_{x \in [0, 1]} |f(x)|$.

Notation. If $\|\cdot\|$ is a norm on V , we say the pair $(V, \|\cdot\|)$ is a *normed space*.

Definition. Suppose $(V, \|\cdot\|)$ is a normed vector space, and (\mathbf{v}_n) is a sequence in V . We say (\mathbf{v}_n) converges to $\mathbf{v} \in V$ if $\forall \epsilon > 0$, $\exists N$ such that $\forall n > N$, $\|\mathbf{v}_n - \mathbf{v}\| < \epsilon$.

Equivalently, $(\mathbf{v}_n) \rightarrow \mathbf{v}$ if and only if $\|\mathbf{v}_n - \mathbf{v}\| \rightarrow 0$ in \mathbb{R} .

Example. Let $V = \mathbb{R}^n$, $\mathbf{v}_k = (v_{k,1}, \dots, v_{k,n})$.

(a) $(\mathbf{v}_k) \rightarrow \mathbf{v}$ with respect to $\|\cdot\|_\infty$

$$\iff \|\mathbf{v}_k - \mathbf{v}\|_\infty \rightarrow 0$$

$$\iff \max\{|v_{k,i} - v_i|\} \rightarrow 0$$

$$\iff |v_{k,i} - v_i| \rightarrow 0 \text{ for all } 1 \leq i \leq n$$

$$\iff v_{k,i} \rightarrow v_i.$$

So sequence converges if and only if every component converges.

(b) $(\mathbf{v}_k) \rightarrow \mathbf{v}$ with respect to $\|\cdot\|_1$

$$\iff \sum_{i=1}^n |v_{k,i} - v_i| \rightarrow 0$$

$$\iff |v_{k,i} - v_i| \rightarrow 0 \text{ for all } 1 \leq i \leq n$$

$$\iff v_{k,i} \rightarrow v_i.$$

Note the two different norms in (a) and (b) give the same notion of convergence.

We set a convention that, when talking about convergence in \mathbb{R}^n without mentioning a norm, then it's with respect to $\|\cdot\|_1$ (or $\|\cdot\|_\infty$ or $\|\cdot\|_2$) (these all give the same notion of convergence).

Example. Let $V = C[0, 1]$,

$$f_n(x) = \begin{cases} 1 - nx & x \in [0, \frac{1}{n}) \\ 0 & x \in [\frac{1}{n}, 1] \end{cases}$$

So

$$\|f_n\|_1 = \int_0^1 |f_n(x)| dx = \frac{1}{2n} \rightarrow 0$$

as $n \rightarrow \infty$. So $(f_n) \rightarrow 0$ with respect to $\|\cdot\|_1$.

On the other hand, $\|f_n\|_\infty = 1 \not\rightarrow 0$, so $(f_n) \not\rightarrow 0$ with respect to $\|\cdot\|_\infty$. Here the two different norms give two different notions of convergence.

1.2 Continuity

Let $(V, \|\cdot\|)$ be a normed vector space.

Recall: If $\mathbf{v}_n \in V$ and $\mathbf{v} \in V$, the sequence $(\mathbf{v}_n) \rightarrow \mathbf{v}$ if for every $\varepsilon > 0$, there exists n such that $\|\mathbf{v}_n - \mathbf{v}\| < \varepsilon$ when $n > N$.

Definition. Suppose V and W are normed spaces, and $f : V \rightarrow W$. We say f is *continuous* if the sequence $(f(\mathbf{v}_n)) \rightarrow f(\mathbf{v})$ in W whenever $(\mathbf{v}_n) \rightarrow \mathbf{v}$ in V .

Example. (1) $f : V \rightarrow \mathbb{R}^n$, $f(\mathbf{v}) = (f_1(\mathbf{v}), \dots, f_n(\mathbf{v}))$. Then f is continuous if and only if f_1, \dots, f_n are all continuous.

(2) $p_i : \mathbb{R}^n \rightarrow \mathbb{R}$ by $p_i(\mathbf{v}) = v_i$. Then p_i is continuous.

(3) $V = C[0, 1]$, $x \in [0, 1]$, $p_x : C[0, 1] \rightarrow \mathbb{R}$ by $p_x(f) = f(x)$ (linear map). Then p_x is continuous with respect to the uniform norm on $C[0, 1]$:

$$\begin{aligned} (f_n) &\rightarrow f \text{ wrt } \|\cdot\|_\infty \\ \iff \max_{y \in [0, 1]} |f_n(x) - f(x)| &\rightarrow 0 \\ \implies |f_n(x) - f(x)| &\rightarrow 0 \\ \implies (f_n(x)) &\rightarrow f(x) \end{aligned}$$

However, p_x is not continuous with respect to $\|\cdot\|_1$ on $C[0, 1]$. See examples in M&T.

So linear maps may not be continuous.

(4) If $f : V_1 \rightarrow V_2$ and $g : V_2 \rightarrow V_3$ are continuous, so is $g \circ f : V_1 \rightarrow V_3$.

(5) $\|\cdot\| : V \rightarrow \mathbb{R}$ is continuous.

Lemma. If $\mathbf{v}, \mathbf{w} \in V$, then $\|\mathbf{w} - \mathbf{v}\| \geq ||\mathbf{w}\| - \|\mathbf{v}\||$.

Proof. Since $\|\mathbf{v}\| + \|\mathbf{w} - \mathbf{v}\| \geq \|\mathbf{w}\|$,

$$\|\mathbf{w} - \mathbf{v}\| \geq \|\mathbf{w}\| - \|\mathbf{v}\|.$$

Similarly, $\|\mathbf{w} - \mathbf{v}\| = \|\mathbf{v} - \mathbf{w}\| \geq \|\mathbf{v}\| - \|\mathbf{w}\|$. So $\|\mathbf{w} - \mathbf{v}\| \geq ||\mathbf{w}\| - \|\mathbf{v}\||$. \square

Now we can prove the 5th example above:

Proof. Let $f(\mathbf{v}) = \|\mathbf{v}\|$. Then if $(\mathbf{v}_n) \rightarrow \mathbf{v}$, $(\|\mathbf{v}_n - \mathbf{v}\|) \rightarrow 0$. But $\|\mathbf{v}_n - \mathbf{v}\| \geq ||\|\mathbf{v}_n\| - \|\mathbf{v}\|| = |f(\mathbf{v}_n) - f(\mathbf{v})| \geq 0$. So by squeeze rule, $(|f(\mathbf{v}_n) - f(\mathbf{v})|) \rightarrow 0$, i.e. $f(\mathbf{v}_n) \rightarrow f(\mathbf{v})$. \square

Proposition. $f : V \rightarrow W$ is continuous if and only if for every $\mathbf{v} \in V$ and $\varepsilon > 0$, there exists $\delta > 0$ such that

$$\|f(\mathbf{w}) - f(\mathbf{v})\|_W < \varepsilon$$

whenever $\|\mathbf{w} - \mathbf{v}\|_V < \delta$.

Proof. Suppose the $\varepsilon - \delta$ condition hold. We'll show that f is continuous, i.e. if $(\mathbf{v}_n) \rightarrow \mathbf{v}$, then $(f(\mathbf{v}_n)) \rightarrow f(\mathbf{v})$.

Given $(\mathbf{v}_n) \rightarrow \mathbf{v}$ and $\varepsilon > 0$, pick $\delta > 0$ such that $\|f(\mathbf{w}) - f(\mathbf{v})\| < \varepsilon$ whenever $\|\mathbf{w} - \mathbf{v}\| < \delta$. Since $(\mathbf{v}_n) \rightarrow \mathbf{v}$, there exists N such that $\|\mathbf{v}_n - \mathbf{v}\| < \delta$ whenever $n > N$, i.e. $\|f(\mathbf{v}_n) - f(\mathbf{v})\| < \varepsilon$ when $n > N$. So $(f(\mathbf{v}_n)) \rightarrow f(\mathbf{v})$. So f is continuous.

If the $\varepsilon - \delta$ condition does not hold, then there exists $\mathbf{v} \in V$ and $\varepsilon > 0$ such that for every $n > 0$, there exists \mathbf{v}_n with

$$\|\mathbf{v} - \mathbf{v}_n\| < \frac{1}{n}$$

but

$$\|f(\mathbf{v}) - f(\mathbf{v}_n)\| > \varepsilon$$

(Otherwise, take $\delta = \frac{1}{n}$ and we get a contradiction). Then $(\mathbf{v}_n) \rightarrow \mathbf{v}$, but $(f(\mathbf{v}_n)) \not\rightarrow f(\mathbf{v})$. So f is not continuous. \square

1.2.1 Addendum

Suppose V, W are normed spaces and U_α is an open subset of V for all $\alpha \in A$. Let $U = \cup_{\alpha \in A} U_\alpha$.

Proposition. Suppose $f : U \rightarrow W$ and f is continuous on all U_α . Then f is continuous on U . It's important that U_α 's are all open. For example, any $f : V \rightarrow W$ is continuous on $\{\mathbf{v}\}$, but may not be continuous on $\cup_{\mathbf{v} \in V} \{\mathbf{v}\} = V$.

Proof. Must show that given $\mathbf{v} \in U$ and $\varepsilon > 0$, $\exists \delta > 0$ s.t.

$$f(B_\delta(\mathbf{v}) \cap U) \subset B_\varepsilon(f(\mathbf{v}))$$

$\mathbf{v} \in \cup_{\alpha \in A} U_\alpha$, so $\mathbf{v} \in U_{\alpha_0}$ for some $\alpha_0 \in A$. f is continuous on U_{α_0} , so $\exists \delta_1 > 0$ s.t.

$$f(B_{\delta_1}(\mathbf{v}) \cap U_{\alpha_0}) \subset B_\varepsilon(f(\mathbf{v}))$$

U_{α_0} is open, so $\exists \delta_2 > 0$ s.t. $B_{\delta_2}(\mathbf{v}) \subset U_{\alpha_0}$.

Let $\delta = \min(\delta_1, \delta_2)$. Then $B_\delta(\mathbf{v}) \subset B_{\delta_1}(\mathbf{v})$ and $B_\delta(\mathbf{v}) \subset B_{\delta_2}(\mathbf{v}) \subset U_{\alpha_0}$.

So $B_\delta(\mathbf{v}) \subset B_{\delta_1}(\mathbf{v}) \cap U_{\alpha_0}$.
Thus

$$f(B_\delta(\mathbf{v}) \cap U) = f(B_\delta(\mathbf{v})) \subset f(B_{\delta_1}(\mathbf{v}) \cap U_{\alpha_0}) \subset B_\varepsilon(f(\mathbf{v}))$$

□

1.3 Open and Closed Subsets

Definition. If $\mathbf{v} \in V$ and $r > 0$,

$$B_r(\mathbf{v}) = \{\mathbf{w} \in V \mid \|\mathbf{v} - \mathbf{w}\| < r\}$$

is the *open ball* of radius r centered at \mathbf{v} ,

$$B_r(\mathbf{v}) = \{\mathbf{w} \in V \mid \|\mathbf{v} - \mathbf{w}\| \leq r\}$$

is the *closed ball* of radius r centered at \mathbf{v} .

Now we can get an alternative definition of continuous:

• f is continuous if and only if for every $\mathbf{v} \in V$ and $\varepsilon > 0$, there exists $\delta > 0$ such that $f(B_\delta(\mathbf{v})) \subset B_\varepsilon(f(\mathbf{v}))$.

Definition. $U \subset V$ is an *open subset* of V if for every $\mathbf{u} \in U$, there exists $\varepsilon > 0$ such that $B_\varepsilon(\mathbf{u}) \subset U$.

Proposition. If $f : V \rightarrow W$ is continuous and $U \subset W$ is open, then $f^{-1}(U)$ is open in V .

Proof. Suppose $\mathbf{v} \in f^{-1}(U)$, i.e. $f(\mathbf{v}) \in U$.

U is open, so there exists $\varepsilon > 0$ such that $B_\varepsilon(f(\mathbf{v})) \subset U$.

f is continuous, so $\exists \delta > 0$ such that $f(B_\delta(\mathbf{v})) \subset B_\varepsilon(f(\mathbf{v})) \subset U$, i.e. $B_\delta(\mathbf{v}) \subset f^{-1}(U)$ so $f^{-1}(U)$ is open.

The converse is also true (see M&T). □

Definition. (Open subsets) Recall $U \subset V$ is *open* in V if for every $\mathbf{u} \in U$, $\exists \varepsilon > 0$ s.t. $B_\varepsilon(\mathbf{u}) \subset U$.

Proposition. If $f : V \rightarrow W$ is continuous and $U \subset W$ is open, then $f^{-1}(U)$ is open in V .

Example. Given $\mathbf{v} \in V$, define

$$\begin{aligned} f_{\mathbf{v}} : V &\rightarrow \mathbb{R} \\ f_{\mathbf{v}}(\mathbf{w}) &= \|\mathbf{v} - \mathbf{w}\| \end{aligned}$$

Then $f_{\mathbf{v}}$ is continuous, so

$$B_r(\mathbf{v}) = f_{\mathbf{v}}^{-1}((-r, r))$$

is open in V , i.e. open balls are open.

Definition. (Closed subsets) Recall if $C \subset V$, $V - C = \{\mathbf{v} \in V | \mathbf{v} \notin C\}$ is the complement of C . $C \subset V$ is closed if $V - C$ is an open subset of V .

Corollary. If $f : V \rightarrow W$ is continuous and C is closed in W , then $f^{-1}(C)$ is closed in V .

Example. Let

$$C = \{(x, f(x)) | x \in \mathbb{R}\}$$

where $f : \mathbb{R} \rightarrow \mathbb{R}$ is continuous. Then C is closed in \mathbb{R}^2 .

Proof. Let $F : \mathbb{R}^2 \rightarrow \mathbb{R}$ by $F(x, y) = f(x) - y$ which is continuous. Then $C = F^{-1}(\{0\})$ is closed, since $\{0\}$ is closed in \mathbb{R} . \square

Example.

$$\overline{B}_r(\mathbf{v}) = f_{\mathbf{v}}^{-1}([0, r])$$

is closed in any normed space V .

Example. $\mathbb{Q} \subset \mathbb{R}$ is neither open nor closed.

Example. $V \subset V$, $\phi \subset V$ are both open and closed.

Proposition. C is closed in V if and only if for every sequence $(\mathbf{v}_n) \rightarrow \mathbf{v} \in V$ which satisfies $\mathbf{v}_n \in C$ for all n , we have $\mathbf{v} \in C$ as well.

Proof. Suppose C is closed in V , and $(\mathbf{v}_n) \rightarrow \mathbf{v}$ with $\mathbf{v} \notin C$. Now $V - C$ is open, and $\mathbf{v} \in V - C$. So $\exists \varepsilon > 0$ s.t. $B_\varepsilon(\mathbf{v}) \subset V - C$. Since $(\mathbf{v}_n) \rightarrow \mathbf{v}$, there exists N s.t. $\mathbf{v}_n \in B_\varepsilon(\mathbf{v}) \subset V - C$ for all $n > N$. So $\mathbf{v}_n \notin C$. Contradiction.

Conversely, suppose that C is not closed. Then $V - C$ is not open. So there exists $\mathbf{u} \in V - C$ such that for every $\varepsilon > 0$, $B_\varepsilon(\mathbf{u}) \not\subset V - C$, i.e. $B_\varepsilon(\mathbf{u}) \cap C \neq \emptyset$. Now pick \mathbf{v}_n s.t. $\mathbf{v}_n \in B_{1/n}(\mathbf{u}) \cap C$. Then $\|\mathbf{v}_n - \mathbf{u}\| < \frac{1}{n} \rightarrow 0$, so $(\mathbf{v}_n) \rightarrow \mathbf{u}$ for all $\mathbf{v}_n \in C$, but $\mathbf{u} \notin C$. Contradiction. \square

1.4 Lipschitz equivalence

We've seen in the first lecture that $\|\cdot\|_1, \|\cdot\|_2, \|\cdot\|_\infty$ all induce the same notion of convergence on \mathbb{R}^n . So $f : \mathbb{R}^n \rightarrow V$ is continuous with respect to $\|\cdot\|$ if and only if it's continuous with respect to $\|\cdot\|_\infty$.

Proposition. Suppose $\|\cdot\|, \|\cdot\|'$ are two norms on V . The map $id : (V, \|\cdot\|) \rightarrow (V, \|\cdot\|')$ by $id(\mathbf{v}) = \mathbf{v}$ is continuous if and only if there exists some constants $C > 0$ such that

$$\|\mathbf{v}\|' \leq C\|\mathbf{v}\|$$

for all $\mathbf{v} \in V$.

Proof. Suppose $\|\mathbf{v}\|' \leq C\|\mathbf{v}\|$ for all $\mathbf{v} \in V$.

If $(\mathbf{v}_n) \rightarrow \mathbf{v}$ with respect to $\|\cdot\|$, then $(\|\mathbf{v} - \mathbf{v}_n\|) \rightarrow 0$. But then

$$0 \leq \|\mathbf{v} - \mathbf{v}_n\|' \leq C\|\mathbf{v} - \mathbf{v}_n\|$$

By the squeeze law, $\|\mathbf{v} - \mathbf{v}_n\|' \rightarrow 0$ as well. So $(\mathbf{v}_n) \rightarrow \mathbf{v}$ with respect to $\|\cdot\|'$. This means $id : (V, \|\cdot\|) \rightarrow (V, \|\cdot\|')$ is continuous.

Conversely, suppose $id : (V, \|\cdot\|) \rightarrow (V, \|\cdot\|')$ is continuous. Then there exists $\delta > 0$ s.t. $B_\delta(\mathbf{0}, \|\cdot\|) \subset B_1(\mathbf{0}, \|\cdot\|')$.

For any $\mathbf{v} \in V, \mathbf{v} \neq \mathbf{0}$, there exists k s.t. $\|k\mathbf{v}\| = \frac{\delta}{2}$. So $k\mathbf{v} \in B_\delta(\mathbf{0}, \|\cdot\|)$, so $k\mathbf{v} \in B_1(\mathbf{0}, \|\cdot\|')$, i.e. $\|k\mathbf{v}\|' < 1 = \frac{2}{\delta}\|k\mathbf{v}\|$. Divide by $|k|$ we get

$$\|\mathbf{v}\|' \leq \frac{2}{\delta}\|\mathbf{v}\|$$

for all $\mathbf{v} \neq \mathbf{0}$. So we can take $C = \frac{2}{\delta}$. The case $\mathbf{v} = \mathbf{0}$ is trivial. \square

Definition. If $\|\cdot\|$ and $\|\cdot\|'$ are two norms on V , we say they are *Lipschitz equivalent* if there exists $C > 0$ s.t.

$$\frac{1}{C}\|\mathbf{v}\| \leq \|\mathbf{v}\|' \leq C\|\mathbf{v}\|$$

for all $\mathbf{v} \in V$, or say there exists C_1, C_2 such that

$$\|\mathbf{v}\| \leq C_1\|\mathbf{v}\|'$$

and

$$\|\mathbf{v}\|' \leq C_2\|\mathbf{v}\|$$

That is also equivalent to

$$id : (V, \|\cdot\|) \rightarrow (V, \|\cdot\|')$$

and

$$id : (V, \|\cdot\|') \rightarrow (V, \|\cdot\|)$$

being both continuous.

Corollary. If $\|\cdot\|$ and $\|\cdot\|'$ are Lipschitz equivalent, then:

- (a) $(\mathbf{v}_n) \rightarrow \mathbf{v}$ with respect to $\|\cdot\|$ if and only if $(\mathbf{v}_n) \rightarrow \mathbf{v}$ with respect to $\|\cdot\|'$.
- (b) $f : V \rightarrow W$ is continuous with respect to $\|\cdot\|$ if and only if $f : V \rightarrow W$ is continuous with respect to $\|\cdot\|'$.
- (c) $g : W \rightarrow V$ is continuous with respect to $\|\cdot\|$ if and only if $g : W \rightarrow V$ is continuous with respect to $\|\cdot\|'$.

Example. $\|\mathbf{v}\|_\infty \leq \|\mathbf{v}\|_2 \leq \|\mathbf{v}\|_1 \leq n\|\mathbf{v}\|_\infty$ for all $\mathbf{v} \in \mathbb{R}^n$. So $\|\cdot\|_\infty, \|\cdot\|_2, \|\cdot\|_1$ are all Lipschitz equivalent.

Problem. Can we find a norm on \mathbb{R}^n that is not Lipschitz equivalent to these?

2 Uniform Convergence

2.1 Notions of Convergence

Let $A \subset \mathbb{R}$, $f, f_n : A \rightarrow \mathbb{R}$.

We've known the definition of continuous and boundedness from Analysis I. Now define $C(A)$ to be the set of continuous functions $f : A \rightarrow \mathbb{R}$, and $B(A)$ to be the set of bounded functions $F : A \rightarrow \mathbb{R}$. Both of these are vector spaces.

We have $C[0, 1] \subset B[0, 1]$ by maximum value theorem, while $C(0, 1) \not\subset B(0, 1)$ (take $f(x) = \frac{1}{x}$).

Definition. If $f, f_n : A \rightarrow \mathbb{R}$, we say $(f_n) \rightarrow f$ *pointwise* if $(f_n(x)) \rightarrow f(x)$ for every $x \in A$.

Definition. The *uniform norm* $\|\cdot\|_\infty$ on $B(A)$ is given by

$$\|f\|_\infty = \sup_{x \in A} |f(x)|$$

If $f, f_n : A \rightarrow \mathbb{R}$, we say $(f_n) \rightarrow f$ *uniformly* if $\|f - f_n\|_\infty \rightarrow 0$.

Equivalently, if $(f_n) \rightarrow f$ pointwise, then for every $x \in A$ and $\epsilon > 0$, $\exists N$ s.t. $|f_n(x) - f(x)| < \epsilon$ whenever $n > N$.

If $(f_n) \rightarrow f$ uniformly, given ϵ , we need to find some N that works for all $x \in A$.

Example. Let $A = \mathbb{R}$, $f_n(x) = x + \frac{1}{n}$, $f(x) = x$. Then $(f_n) \rightarrow f$ pointwise and uniformly.

Example. Let $A = \mathbb{R}$, $g_n(x) = (x + \frac{1}{n})^2$, $g(x) = x^2$. Then $g(n) \rightarrow g$ pointwise, but $g_n - g = \frac{2x}{n} + \frac{1}{n^2}$ is not even bounded. So (g_n) does not converge to g uniformly. Nevertheless, $(g_n) \rightarrow g$ uniformly on $[a, b]$ for any $a, b \in \mathbb{R}$ (since convergence and uniform convergence is the same on compact sets).

Example. If $(f_n) \rightarrow f$ uniformly, then $(f_n) \rightarrow f$ pointwise (Immediate from definition).

Theorem. Suppose $f_n \in C(A)$ and $(f_n) \rightarrow f$ uniformly on A . Then $f \in C(A)$.

Proof. Given $x \in A$ and $\epsilon > 0$, we need to find $\delta > 0$ s.t.

$$|f(x) - f(y)| < \epsilon$$

whenever $|x - y| < \delta$ and $y \in A$.

Since $(f_n) \rightarrow f$ uniformly, $\exists N$ s.t.

$$|f_n(y) - f(y)| < \frac{\epsilon}{4}$$

whenever $n \geq N$ and $y \in A$.

Since f_N is continuous, $\exists \delta > 0$ s.t.

$$|f_N(x) - f_N(y)| < \frac{\epsilon}{2}$$

whenever $|x - y| < \delta$ and $y \in A$. Then for $|x - y| < \delta$ and $y \in A$,

$$\begin{aligned} |f(x) - f(y)| &\leq |f(x) - f_N(x)| + |f_N(x) - f_N(y)| + |f_N(y) - f(y)| \\ &< \frac{\varepsilon}{4} + \frac{\varepsilon}{2} + \frac{\varepsilon}{4} = \varepsilon \end{aligned}$$

which is what we wanted to prove. \square

Corollary. $C[a, b]$ is a closed subset of $B[a, b]$ with respect to $\|\cdot\|_\infty$.

Proof. Recall that C is closed if $c \in C$ whenever $(c_n) \rightarrow c$ and $c_n \in C$. \square

Example. Let $A = [0, 1]$, $f_n(x) = x^n$, $f(x) = \begin{cases} 0 & x \in [0, 1) \\ 1 & x = 1 \end{cases}$.

Then $(f_n) \rightarrow f$ pointwise but not uniformly, since $f_n \in C[0, 1]$, but $f \notin C[0, 1]$.

Example. Let $f_n(x) = (1 - x)x^n$. Then $(f_n) \rightarrow 0$ pointwise. In fact $(f_n) \rightarrow 0$ uniformly.

Proof. Given $\varepsilon > 0$, we must find N s.t. $|f_n(x)| < \varepsilon$ for all $x \in [0, 1]$ whenever $n > N$.

We know $1 - \varepsilon < 1$, so $(1 - \varepsilon)^n \rightarrow 0$. Pick N s.t. $(1 - \varepsilon)^n < \varepsilon$ whenever $n > N$. Then for $n > N$,

$$|(1 - x)x^n| < 1 \cdot (1 - \varepsilon)^n < \varepsilon$$

for $x \in [0, 1 - \varepsilon]$, and

$$|(1 - x)x^n| < \varepsilon \cdot 1^n = \varepsilon$$

for $x \in (1 - \varepsilon, 1]$. \square

Everything so far in this chapter works for $f : A \rightarrow W$, where $A \subset V$ and V, W are both normed spaces. (exercise)

Recall that if $f, f_n \in C[a, b]$ with $a, b \in \mathbb{R}$, then $(f_n) \rightarrow f$ in L^1 (with respect to $\|\cdot\|_1$) if

$$\|f_n - f\|_1 = \int_a^b |f_n(x) - f(x)| dx \rightarrow 0$$

Lemma. If $(f_n) \rightarrow f$ uniformly on $[a, b]$ and $f_n \in C[a, b]$, then $(f_n) \rightarrow f$ in L^1 on $[a, b]$.

Proof. $(f_n) \rightarrow f$ uniformly implies that $f \in C[a, b]$. Given $\varepsilon > 0$, pick N s.t.

$$|f_n(x) - f(x)| < \frac{\varepsilon}{(b - a)}$$

for $n > N$ and $x \in [a, b]$. Then

$$\|f_n - f\|_1 = \int_a^b |f_n(x) - f(x)| dx < \int_a^b \frac{\varepsilon}{b - a} dx = \varepsilon$$

So $(f_n) \rightarrow f$ in L^1 . \square

Example. Let $A = [0, 1]$,

$$f_n(x) = \begin{cases} nx & x \in \left[0, \frac{1}{n}\right] \\ 2 - nx & x \in \left[\frac{1}{n}, \frac{2}{n}\right] \\ 0 & x \in \left[\frac{2}{n}, 1\right] \end{cases}$$

Then $(f_n) \rightarrow 0$ pointwise, and in L^1 , but not uniformly.

Example. Let $A = [0, 1]$,

$$f_n(x) = \begin{cases} n^2x & x \in \left[0, \frac{1}{n}\right] \\ 2n - n^2x & x \in \left[\frac{1}{n}, \frac{2}{n}\right] \\ 0 & x \in \left[\frac{2}{n}, 1\right] \end{cases}$$

Then $(f_n) \rightarrow f$ pointwise, but not in L^1 , nor uniformly.

We would like to say that a sequence of bounded integrable functions on $[0, 1]$ that converges pointwise converges in L^1 . But for this to be true, we need a better definition of \int (in measure and probability).

2.2 Power series

Recall some facts about series of complex numbers from Analysis I, for $\sum_{i=0}^{\infty} c_i$, $c_i \in \mathbb{C}$:

- 1) $\sum_{i=0}^{\infty} c_i = c$ means $(\sum_{i=0}^n c_i) \rightarrow c$;
- 2) $\sum_{i=0}^{\infty} c_i$ converges if and only if $\sum_{i=k}^{\infty} c_i$ converges;
- 3) $\sum_{i=k}^{\infty} \alpha^i = \frac{\alpha^k}{1-\alpha}$ if $|\alpha| < 1$;
- 4) If $\sum_{i=0}^{\infty} c_i$ converges, then $(c_n) \rightarrow 0$;
- 5) If $0 < a_i < b_i$ for all i (here $a_i, b_i \in \mathbb{R}$), and $\sum_{i=0}^{\infty} b_i$ converges, then $\sum_{i=0}^{\infty} a_i$ converges as well;
- 6) If $\sum_{i=0}^{\infty} |c_i|$ converges, then $\sum_{i=0}^{\infty} c_i$ converges.

Corollary. If $|c_i| < b_i$ for all i and $\sum_{i=0}^{\infty} b_i$ converges, then $\sum_{i=0}^{\infty} c_i$ converges.

Proof. Follows from (5) and (6). □

Definition. A *power series* is

$$\sum_{i=0}^{\infty} a_i (z_i)^i$$

where $a_i, c, z \in \mathbb{C}$. Call c the *center* of the series.

Proposition. Suppose $\sum_{i=0}^{\infty} a_i (z_0 - c)^i$ converges for some $z_0 \in \mathbb{C}$. Then the series $\sum_{i=0}^{\infty} a_i (z_0 - c)^i$ converges for all z with $|z - c| < |z_0 - c|$.

Proof. By (4), $(a_i (z_0 - c)^i) \rightarrow 0$. Pick N such that $|a_i (z_0 - c)^i| < 1$ for all $i \geq N$.

By (2), suffices to show that $\sum_{i=N}^{\infty} a_i (z - c)^i$ converges. Now

$$|a_i (z - c)^i| = |a_i (z_0 - c)^i| \cdot \left| \frac{z - c}{z_0 - c} \right|^i \leq 1 \cdot \alpha^i$$

(call this 'Key Estimate', to be used later) for $i \geq N$ where $\alpha = \left| \frac{z - c}{z_0 - c} \right|$.

For $|z - c| < |z_0 - c|$, $\alpha < 1$, so $\sum_{i=N}^{\infty} \alpha^i$ converges.

By corollary, it follows that $\sum_{i=0}^{\infty} a_i (z - c)^i$ converges. \square

Definition.

$$R = \sup \left\{ |z - c| \mid \sum_{i=0}^{\infty} a_i (z - c)^i \text{ converges} \right\}$$

is the *radius of convergence* of this series.

The above proposition says that $\sum_{i=0}^{\infty} a_i (z - c)^i$ converges for all $z \in B_R(c) = \{z \in \mathbb{C} \mid |z - c| < R\}$.

We can define $f : B_R(c) \rightarrow \mathbb{C}$ by

$$f(z) = \sum_{i=0}^{\infty} a_i (z - c)^i$$

Let

$$p_n(z) = a_i (z - c)^i$$

Then $(p_n) \rightarrow f$ pointwise on $B_R(c)$.

Theorem. With notation as above, $(p_n) \rightarrow f$ uniformly on $\bar{B}_r(c) = \{z \in \mathbb{C} \mid |z - c| \leq r\}$ for any $r < R$.

Proof. Fix $z_0 \in \mathbb{C}$ with $r < |z_0 - c| < R$. Then $\sum_{i=0}^{\infty} a_i (z_0 - c)^i$ converges. Let

$$E_n(z) = f(z) - p_n(z) = \sum_{i=n+1}^{\infty} a_i (z - c)^i$$

We want to show that given $\varepsilon > 0$, $\exists N$ s.t. $|E_n(z)| < \varepsilon$ for all $n > N$ and $z \in \bar{B}_r(c)$.

Pick N_0 with $|a_i (z_0 - c)^i| < 1$ for all $i \geq N_0$ as in the proof of the previous proposition.

Now for $n > N_0$, Key Estimate says that

$$\begin{aligned} |E_n(z)| &= \left| \sum_{i=n+1}^{\infty} a_i (z - c)^i \right| \\ &\leq \sum_{i=n+1}^{\infty} |a_i (z - c)^i| \\ &\leq \sum_{i=n+1}^{\infty} \alpha(z)^i \end{aligned}$$

where $\alpha(z) = \frac{|z-c|}{|z_0-c|}$.

If $z \in \bar{B}_r(c)$, $\alpha(z) \leq \alpha_0 = \frac{r}{|z_0-c|} < 1$. So

$$|E_n(z)| \leq \sum_{i=1}^{\infty} \alpha^i = \frac{\alpha_0^{n+1}}{1-\alpha_0}$$

Now $\alpha_0 < 1$, so $\frac{\alpha_0^{n+1}}{1-\alpha_0} \rightarrow 0$ as $n \rightarrow \infty$. Pick $N > N_0$ s.t. $\frac{\alpha_0^{n+1}}{1-\alpha_0} < \varepsilon$ for $n > N$. Then $|E_n(z)| < \varepsilon$ for all $n > N$ and $z \in \bar{B}_r(c)$ which is what we wanted. \square

Remark. (p_n) may not converge uniformly on $B_R(c)$. For example, $\sum_{i=0}^{\infty} x^i$ has $R = 1$, and equals $f(x) = \frac{1}{1-x}$ on $B_1(0)$, but p_n is a polynomial, so bounded on $\bar{B}_1(0)$, so $f(x) - p_n(x)$ is not even a bounded function on $B_1(0)$.

Corollary.

$$f(z) = \sum_{i=0}^{\infty} a_i (z-c)^i$$

is a continuous map $f : B_R(c) \rightarrow \mathbb{C}$.

Proof. $p_n = \sum_{i=0}^n a_i (z-c)^i$ is a polynomial, so is continuous as a map $\mathbb{C} \rightarrow \mathbb{C}$. $(p_n) \rightarrow f$ uniformly on $\bar{B}_r(c)$ for any $r < R$, so $f : \bar{B}_r(c) \rightarrow \mathbb{C}$ is continuous for any $r < R$.

Given $z \in B_R(c)$, pick r with $z \in B_r(c)$. Then f is continuous at z . So f is continuous at all $z \in B_R(c)$, i.e. $f : B_R(c) \rightarrow \mathbb{C}$ is continuous. \square

We can now construct lots of continuous functions using power series.

Example.

$$\exp(z) = \sum_{i=0}^{\infty} \frac{z^i}{i!}$$

has $R = \infty$, so is a well defined, continuous function on \mathbb{C} .

Let $f(x) = \exp(x)$ for $x \in \mathbb{R}$. We want to show that $f'(x) = f(x)$:

$$\frac{d}{dx} \left(\sum_{i=0}^{\infty} \frac{x^i}{i!} \right) = \sum_{i=0}^{\infty} \frac{ix^{i-1}}{i!} = \sum_{i=1}^{\infty} \frac{x^{i-1}}{(i-1)!} = \exp(x)$$

this looks easy, but why does the first equality hold?

Example. Suppose

$$\sum_{i=0}^{\infty} a_i (z-c)$$

has radius of convergence R . Then if $p_n = \sum_{i=0}^n a_i (z-c)^i$, $(p_n) \rightarrow f(z) = \sum_{i=0}^{\infty} a_i (z-c)^i$ uniformly on $\bar{B}_r(c)$ for all $r < R \implies f$ is continuous on $\bar{B}_r(c)$ for $r < R$.

Take $U_r = B_r(c)$, so f is continuous on U_r for $r < R$. U_r is open. So f is continuous on $\cup_{r < R} U_r = B_R(c)$.

2.3 Integration and Differentiation

Recall from Analysis I:

Theorem. (Fundamental Theorem of Calculus) If $f \in C[a, b]$, then

$$F(x) = \int_{x_0}^x f(y) dy$$

exists, and

$$F'(x) = f(x).$$

Some properties of integral:

Suppose $f, g \in C[a, b]$.

(1)

$$\int_{x_0}^x f(y) + \lambda g(y) dy = \int_{x_0}^x f(y) dy + \lambda \int_{x_0}^x g(y) dy$$

(2) If $f(y) \leq g(y)$ for all $y \in [a, b]$, then

$$\int_{x_0}^x f(y) dy \leq \int_{x_0}^x g(y) dy$$

(3)

$$\left| \int_x^{x_0} f(y) dy \right| \leq \left| \int_x^{x_0} |f(y)| dy \right|$$

Suppose $f_n \in C[a, b]$ and $(f_n) \rightarrow f$ uniformly on $[a, b]$. So $f \in C[a, b]$. Thus

$$F(x) = \int_{x_0}^x f_n(y) dy$$

and

$$F(x) = \int_{x_0}^x f(y) dy$$

are defined.

Proposition. $(F_n) \rightarrow F$ uniformly on $[a, b]$.

Proof. $(f_n) \rightarrow f$ uniformly, so given $\varepsilon > 0$, $\exists N$ s.t.

$$|f_n(x) - f(x)| < \varepsilon$$

for all $n > N$ and $x \in [a, b]$. Choose N s.t.

$$|f_n(x) - f(x)| < \frac{\varepsilon}{b-a}$$

for all $n > N$ and $x \in [a, b]$. Then for $x \in [a, b]$,

$$\begin{aligned} |F_n(x) - F(x)| &= \left| \int_{x_0}^x (f_n(y) - f(y)) dy \right| \\ &\leq \left| \int_{x_0}^x |f_n(y) - f(y)| dy \right| \\ &\leq \left| \int_{x_0}^x \frac{\varepsilon}{b-a} dy \right| dy \\ &= \frac{\varepsilon |x - x_0|}{|b-a|} \\ &\leq \varepsilon \end{aligned}$$

So $(F_n) \rightarrow F$ uniformly on $[a, b]$. \square

Note that $(f_n) \in C(\mathbb{R})$, $(f_n) \rightarrow f$ uniformly does not imply $(F_n) \rightarrow F$ uniformly on \mathbb{R} . (But does on $[a, b]$ for $a, b \in \mathbb{R}$).

Let

$$f(y) = \sum_{i=0}^{\infty} a_i (y - c)^i$$

be a real power series $(a_i, c, y \in \mathbb{R})$ with radius of convergence R . Then if the partial sum $p_n(y) = \sum_{i=0}^n a_i (y - c)^i$, then $(p_n) \rightarrow f$ uniformly on $[c - r, c + r]$ for any $r < R$.

Corollary.

$$\int_c^x f(y) dy = \sum_{i=0}^{\infty} \frac{a_i}{i+1} (x - c)^{i+1}$$

for all $x \in (c - R, c + R)$.

Proof. Given $x \in (c - R, c + R)$, pick r with $|x - c| < r < R$. Then $(p_n) \rightarrow f$ uniformly on $[c - r, c + r]$, so by proposition

$$(P_n) \rightarrow \int_c^x f(y) dy$$

where

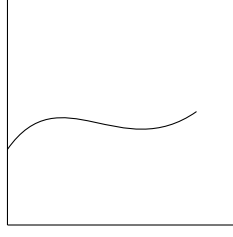
$$P_n = \int_c^x p_n(y) dy = \sum_{i=0}^n \frac{a_i}{i+1} (x - c)^{i+1}$$

\square

Q: If $(f_n) \rightarrow f$ uniformly, what can I say about (f_n) ?

A: Nothing, because:

Example. Take $f_n(x) = \frac{1}{n} \sin nx$, $x \in [0, \pi]$. Then $(f_n) \rightarrow 0$ uniformly on $[0, \pi]$, but $f'_n(x) = \cos nx$ doesn't converge for any $x \in (0, \pi)$.



Proposition. If

$$f(y) = \sum_{i=0}^{\infty} a_i (y - c)^i$$

converges on $(c - R, c + R)$, then

$$f'(y) = \sum_{i=0}^{\infty} i a_i (y - c)^{i-1}$$

on $(c - R, c + R)$.

Proof.

Lemma.

$$\sum_{i=0}^{\infty} i a_i (y - c)^{i-1}$$

converges for all $y \in (c - R, c + R)$.

Pick y_0 with $|y - c| < |y_0 - c| < R$.

$\sum_{i=0}^{\infty} a_i (y - c)^i$ converges, so by 'Key Estimate', $\exists N$ s.t.

$$|a_i (y - c)|^i < \alpha^i$$

for all $i \geq N$, where $\alpha = \left| \frac{y-c}{y_0-c} \right| < 1$.

If $y = c$, $\sum i a_i (y - c)^{i-1}$ obviously converges. If not, estimate

$$\left| i a_i (y - c)^{i-1} \right| < \frac{i}{|y - c|} \alpha^i$$

Now $\sum_{i=0}^{\infty} \frac{i}{|y-c|} \alpha^i$ converges by Ratio Test. So $\sum_{i=0}^{\infty} i a_i (y - c)^{i-1}$ converges as well. \square

Now begin the proof of proposition:

$$g(y) = \sum_{i=0}^{\infty} i a_i (y - c)^{i-1}$$

is continuous on $(c - R, c + R)$. So by corollary,

$$\int_c^x g(y) dy = \sum_{i=1}^{\infty} a_i (x - c)^i = f(x) - f(c)$$

By Fundamental Theorem of Calculus, $f'(x) = g(x)$. □

Application: Power series solutions of ODEs are legit (as long as we check the radius of convergence).

3 Compactness

3.1 Compact subsets of \mathbb{R}^n

Let V be a normed space. Then if $(\mathbf{v}_n) \rightarrow \mathbf{v} \in V$ and (\mathbf{v}_{n_j}) is a subsequence of (\mathbf{v}_n) , then $(\mathbf{v}_{n_j}) \rightarrow \mathbf{v}$. We leave this as an exercise.

Definition. $A \subset V$ is bounded if $\exists M \in \mathbb{R}$ s.t. $\|\mathbf{v}\| \leq M$ for all $\mathbf{v} \in A$.

If $\|\cdot\|$ and $\|\cdot\|'$ are Lipschitz equivalent, then boundedness with respect to the two norms are equivalent.

Corollary. (Bolzano-Weierstrass in \mathbb{R}^n) If (\mathbf{v}_k) is a bounded sequence in \mathbb{R}^n , it has a converging subsequence.

Proof. To prove this, simply pick a subsequence with the first coordinate convergent, then pick a subsequence of that subsequence with the second coordinate convergent, etc..

Let $\mathbf{v}_k = (v_{1,k}, \dots, v_{n,k})$.

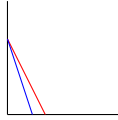
(\mathbf{v}_k) is bounded, so $(v_{i,k})$ is bounded for all $1 \leq i \leq n$. By B-W theorem, there exists a convergent subsequence (v_{1,k_j^1}) of $(v_{1,k})$. Now the sequence (v_{2,k_j^1}) is bounded. So by B-W, there exists a subsequence (v_{2,k_j^2}) which converges. Then by the previous exercise, (v_{1,k_j^2}) converges.

Now consider the sequence (v_{3,k_j^2}) . By B-W, it has a convergent subsequence (v_{3,k_j^3}) . etc.

Apply B-W n times, we get $(\mathbf{v}_{k_j^n})$ of original (\mathbf{v}_n) s.t. (v_{i,k_j^n}) converges for $1 \leq i \leq n$. So $(\mathbf{v}_{k_j^n})$ converges. \square

Example. Let $V = C[0, 1]$ with $\|\cdot\|_\infty$, and

$$f_n(x) = \begin{cases} 1 - nx & x \in [0, \frac{1}{n}] \\ 0 & x \in [\frac{1}{n}, 1] \end{cases}$$



If

$$f(x) = \begin{cases} 1 & x = 0 \\ 0 & x > 0 \end{cases}$$

then $(f_n) \rightarrow f$ pointwise. Then (f_n) is bounded with respect to $\|\cdot\|_\infty$ but has no convergent subsequence.

Proof. Suppose $(f_{n_j}) \rightarrow g$ uniformly, then $(f_{n_j}) \rightarrow g$ pointwise, so $g = f$. But $f \notin C[0, 1]$, so $(f_{n_j}) \not\rightarrow f$ uniformly. \square

Definition. We say $A \subset V$ is sequentially compact (s.compact) if any sequence (\mathbf{v}_n) in A has a convergent subsequence $(\mathbf{v}_{n_j}) \rightarrow \mathbf{v} \in A$.

Example. R is not s.compact, since (n) has no convergent subsequence.

Example. $A = (0, 2)$ is not s.compact, since $(\frac{1}{n}) \rightarrow 0 \notin A$.

Proposition. Suppose $A \subset V$ is s.compact. Then A is closed in V and bounded.

Proof. We prove the contrapositive:

If A is not closed, then there exists a sequence $(\mathbf{v}_n) \rightarrow \mathbf{v}$ with $\mathbf{v}_n \in A$ for all n but $\mathbf{v} \notin A$. By the exercise, any subsequence (\mathbf{v}_{n_j}) converges to $\mathbf{v} \notin A$. So A is not s.compact.

If A is not bounded, then for all $n \in \mathbb{N}$ we can find $\mathbf{v}_n \in A$ with $\|\mathbf{v}_n\| \geq n$. We claim that (\mathbf{v}_{n_j}) has no convergent subsequence: if $(\mathbf{v}_{n_j}) \rightarrow \mathbf{v}$, then $\exists J$ s.t. $\|\mathbf{v}_{n_j} - \mathbf{v}\| < 1$ for all $j > J$. So

$$\|v_{n_j}\| \leq \|\mathbf{v}\| + \|\mathbf{v}_{n_j} - \mathbf{v}\| \leq \|\mathbf{v}\| + 1$$

for all $j > J$, but this is impossible since $n_j \geq j$, so $\|v_{n_j}\| \geq j \rightarrow \infty$ as $j \rightarrow \infty$.

It follows that \mathbf{v}_n has no convergent subsequence, so A is not s.compact. \square

Theorem. (Heine-Borel) $A \subset \mathbb{R}^n$ is s.compact if and only if A is closed and bounded.

Proof. By the proposition, A is s.compact $\implies A$ is closed and bounded. Conversely, suppose A is closed and bounded, and (\mathbf{v}_n) is a sequence in A . Then (\mathbf{v}_n) is bounded (since A is). So by B-W, it has a convergent subsequence. Since A is closed, $\mathbf{v} \in A$. So A is s.compact. \square

Remark. By previous example, $\bar{B}_1(0)$ in $C[0, 1]$ with $\|\cdot\|_\infty$ is closed and bounded but not s.compact since (f_n) has no convergent subsequence. So Heine-Borel theorem does not hold in general spaces.

Remark. If $A \subset V$ a normed space, then A is s.compact $\iff A$ is compact.

Proposition. Suppose $C \subset V$ is s.compact and $f : C \rightarrow W$ is continuous. Then $f(C)$ is s.compact.

Proof. Suppose (\mathbf{w}_n) is a sequence in $f(C)$. Pick $\mathbf{v}_n \in C$ with $f(\mathbf{v}_n) = \mathbf{w}_n$. We know C is s.compact, so (\mathbf{v}_n) has a convergent subsequence $(\mathbf{v}_{n_j}) \rightarrow \mathbf{v} \in C$.

Now f is continuous, so $(\mathbf{w}_{n_j}) = (f(\mathbf{v}_{n_j})) \rightarrow (f(\mathbf{v})) \in f(C)$. So $f(C)$ is s.compact. \square

We'll use the above to prove maximum value theorem.

Lemma. If $A \subset \mathbb{R}$ is closed and bounded, then $\sup A \in A$.

Proof. A is bounded, so $\sup A$ exists. Pick $x_n \in A$ with $\sup A - \frac{1}{n} \leq x_n \leq \sup A$. Then $(x_n) \rightarrow \sup A$. The result follows since A is closed. \square

Theorem. (Maximum value theorem) Suppose C is s.compact, $f : C \rightarrow \mathbb{R}$ is continuous. Then there exists $\mathbf{v} \in C$ s.t.

$$f(\mathbf{v}) \geq f(\mathbf{v}')$$

for all $\mathbf{v}' \in C$.

Proof. We know $A = f(C)$ is a s.compact subset of \mathbb{R} , so it is closed and bounded. So by the lemma, $\sup A$ is in $A = f(C)$. So pick $\mathbf{v} \in C$ with $f(\mathbf{v}) = \sup A$. \square

Application: Norms on \mathbb{R}^n :

Let $\|\cdot\|$ be a norm on \mathbb{R}^n .

Lemma. The map $\text{id} : (\mathbb{R}^n, \|\cdot\|_1) \rightarrow (\mathbb{R}^n, \|\cdot\|)$ is continuous.

Proof. Write $\mathbf{v} = (v_1, \dots, v_n) = \sum_{i=1}^n v_i \mathbf{e}_i$. By the triangle inequality,

$$\|\mathbf{v}\| \leq \sum_{i=1}^n \|v_i \mathbf{e}_i\| = \sum_{i=1}^n |v_i| \|\mathbf{e}_i\| \leq C \sum_{i=1}^n |v_i| = C \|\mathbf{v}\|_1$$

Where $C = \max_{1 \leq i \leq n} \|\mathbf{e}_i\|$. By criterion of section 1.4, the given map is continuous. \square

Corollary. The map $f : (\mathbb{R}^n, \|\cdot\|_1) \rightarrow \mathbb{R}$ given by $f(\mathbf{v}) = \|\mathbf{v}\|$ is continuous.

Theorem. $\|\cdot\|$ is Lipschitz equivalent to $\|\cdot\|_1$.

Proof. Let $S = \{\mathbf{v} \in \mathbb{R}^n \mid \|\mathbf{v}\|_1 = 1\} = g^{-1}(\{1\})$, where $g(\mathbf{v}) = \|\mathbf{v}\|_1$.

Now $g : (\mathbb{R}^n, \|\cdot\|_1) \rightarrow \mathbb{R}$ is continuous, $\{1\}$ is closed in \mathbb{R} , so $g^{-1}(\{1\})$ is closed in $(\mathbb{R}^n, \|\cdot\|_1)$. S is also obviously bounded in $(\mathbb{R}^n, \|\cdot\|_1)$. So S is s.compact by Heine-Borel.

$f : (\mathbb{R}^n, \|\cdot\|_1) \rightarrow \mathbb{R}$, $f(\mathbf{v}) = \|\mathbf{v}\|$ is continuous by corollary. So by maximum value theorem, there exists $\mathbf{v}_{\pm} \in S$ s.t.

$$C_- = f(\mathbf{v}_-) \leq f(\mathbf{v}) \leq f(\mathbf{v}_+) = C_+$$

for all $\mathbf{v} \in S$, i.e. $C_- \leq \|\mathbf{v}\| \leq C_+$ for all $\mathbf{v} \in S$ where $C_- = \|\mathbf{v}_-\| > 0$ since $\mathbf{v}_- \in S \implies \mathbf{v}_- \neq \mathbf{0} \implies \mathbf{v}_- \neq \mathbf{0}$.

Then for $\mathbf{v} \neq \mathbf{0}$ in \mathbb{R}^n , $\mathbf{v}/\|\mathbf{v}\|_1 \in S$. So

$$0 < C_- \leq \left\| \frac{\mathbf{v}}{\|\mathbf{v}\|_1} \right\| \leq C_+$$

i.e.

$$C_- \|\mathbf{v}\|_1 \leq \|\mathbf{v}\| \leq C_+ \|\mathbf{v}\|_1$$

where $C_-, C_+ > 0$. So the two norms are Lipschitz equivalent. \square

Corollary. Any two norms on \mathbb{R}^n are Lipschitz equivalent.

3.2 Completeness

Let V be a normed space, and let (\mathbf{v}_n) be a sequence in V .

Definition. The sequence $(\mathbf{v})_n$ is *Cauchy* if given $\varepsilon > 0$, there exists N s.t. $\|\mathbf{v}_n - \mathbf{v}_m\| < \varepsilon$ for all $n, m \geq N$.

Example. If $(\mathbf{v}_n) \rightarrow \mathbf{v}$, then (\mathbf{v}_n) is Cauchy.

Proof. Given $\varepsilon > 0$, pick N s.t. $\|\mathbf{v}_n - \mathbf{v}\| < \frac{\varepsilon}{2}$ for all $n \geq N$. Then for $n, m \geq N$, by triangle inequality,

$$\|\mathbf{v}_n - \mathbf{v}_m\| \leq \|\mathbf{v}_n - \mathbf{v}\| + \|\mathbf{v} - \mathbf{v}_m\| < \varepsilon$$

i.e. (\mathbf{v}_n) is Cauchy. \square

Example. Let $s_n = \sum_{i=1}^n \frac{1}{i}$. Then s_n diverges. Also it is not Cauchy, even though $|s_n - s_{n+1}| \rightarrow 0$ as $n \rightarrow \infty$.

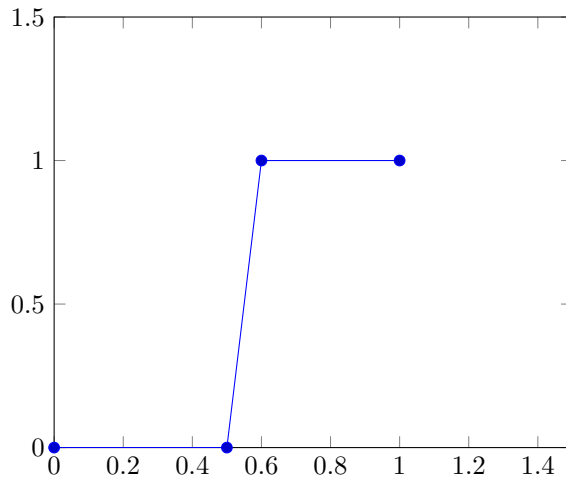
Cauchy sequences *want* to converge.

Example. Given $\varepsilon > 0$, pick N s.t. $\|\mathbf{v}_n - \mathbf{v}_m\| < \varepsilon$ for all $n, m \geq N$. Then all but finitely many terms of (\mathbf{v}_n) are contained in $B_\varepsilon(\mathbf{v}_N)$.

However they may not have an element of V to converge to.

Example. Let $V = C[0, 1]$ with $\|\cdot\|_1$. Take

$$f_n = \begin{cases} 0 & x \in [0, 1/2] \\ n(x - 1/2) & x \in [1/2, 1/2 + 1/n] \\ 1 & x \in [1/2 + 1/n, 1] \end{cases}$$



f_n is Cauchy:

If $m, n \gg N$, $|f_n(x) - f_m(x)| = 0$ if $x \notin A_n = [1/2, 1/2 + 1/N]$, and < 1 if $x \in A_N$. Then

$$\|f_n - f_m\|_1 = \int_0^1 |f_n(x) - f_m(x)| dx \leq \int_{1/2}^{1/2+1/N} 1 dx = \frac{1}{N}$$

so (f_n) is Cauchy.

Now let

$$f(x) = \begin{cases} 0 & x \in [0, 1/2] \\ 1 & x \in (1/2, 1] \end{cases}$$

which is not in $C[0, 1]$.

If $(f_n) \rightarrow g \in C[0, 1]$ then $(f_n) \rightarrow g$ with respect to $\|\cdot\|_1$ on $[0, 1] - A_n$ for any $N > 0$. On the other hand, $(f_n) \rightarrow f$ uniformly on $[0, 1] - A_N$ for any $N > 0$.

On the other hand, $(f_n) \rightarrow f$ uniformly on $[0, 1] - A_N$ for any $N > 0$. So $(f_n) \rightarrow f$ with respect to $\|\cdot\|_1$ on $[0, 1] - A_N$ for all $N > 0$. Therefore $g(x) = f(x)$ for all $x \in [0, 1]$. Contradiction.

Definition. A normed space V is *complete* if every Cauchy sequence (\mathbf{v}_n) in V converges to a limit $\mathbf{v} \in V$.

Example. $(C[0, 1], \|\cdot\|_1)$ is not complete.

Application: Completeness of \mathbb{R}^n .

Let V be a normed vector space, and suppose (\mathbf{v}_n) is a Cauchy sequence in V .

Lemma. (\mathbf{v}_n) is bounded. (Exercise)

Lemma. If (\mathbf{v}_n) has a convergent subsequence $(\mathbf{v}_{n_i}) \rightarrow \mathbf{v} \in V$, then $(\mathbf{v}_n) \rightarrow \mathbf{v}$.

Proof. Given $\varepsilon > 0$, pick M s.t. $\|\mathbf{v}_n - \mathbf{v}_m\| < \frac{\varepsilon}{2}$ whenever $n, m > M$. Now \mathbf{v}_{n_i} converges to \mathbf{v} , so pick I s.t. $\|\mathbf{v}_{n_i} - \mathbf{v}\| < \frac{\varepsilon}{2}$ whenever $i > I$. So choose $I' > I$ s.t. $n_{I'} \geq M$. Then for $n > n_{I'}$,

$$\|\mathbf{v}_n - \mathbf{v}\| \leq \|\mathbf{v}_n - \mathbf{v}_{n_{I'}}\| + \|\mathbf{v}_{n_{I'}} - \mathbf{v}\| < \varepsilon$$

So $(\mathbf{v}_n) \rightarrow \mathbf{v}$. □

Theorem. \mathbb{R}^n is complete.

Proof. Suppose (\mathbf{v}_n) is a Cauchy sequence in \mathbb{R}^n . By lemma 1, (\mathbf{v}_n) is bounded. By B-W, (\mathbf{v}_n) has a convergent subsequence $(\mathbf{v}_{n_i}) \rightarrow \mathbf{v}$. By lemma 2, $(\mathbf{v}_n) \rightarrow \mathbf{v}$, i.e. every Cauchy sequence converges. So \mathbb{R}^n is complete. □

Remark. If $\|\cdot\|$ and $\|\cdot\|'$ are Lipschitz equivalent, then (\mathbf{v}_n) is Cauchy with respect to the two norms are equivalent. So Completeness with respect to the two norms are equivalent.

Since all norms on \mathbb{R}^n are Lipschitz equivalent, the the theorem holds for any norm.

We saw $(C[0, 1], \|\cdot\|_1)$ is not complete. What about $(C[0, 1], \|\cdot\|_\infty)$?

Bounded sequences need not have convergent subsequences.

Theorem. $C[0, 1]$ is complete with respect to $\|\cdot\|_\infty$.

Proof. Given a Cauchy sequence (f_n) , we must find $f \in C[0, 1]$ s.t. $(f_n) \rightarrow f$ uniformly.

Given $\varepsilon > 0$, choose N s.t. $\|f_n - f_m\| < \varepsilon/2$ for all $n, m \geq N$. Then if $x \in [0, 1]$,

$$\begin{aligned} |f_n(x) - f_m(x)| &\leq \max_{x \in [0, 1]} |f_n(x) - f_m(x)| \\ &= \|f_n - f_m\|_\infty \\ &< \varepsilon/2 < \varepsilon \end{aligned}$$

For $n, m \geq N$.

So $(f_n(x))$ is a Cauchy sequence in \mathbb{R} . But \mathbb{R} is complete. So $\lim_{n \rightarrow \infty} f_n(x)$ exists.

Define $f(x) = \lim_{n \rightarrow \infty} f_n(x)$. Then $(f_n) \rightarrow f$ pointwise.

Now we want to prove $(f_n) \rightarrow f$ uniformly. Given $\varepsilon > 0$, and $x \in [0, 1]$, pick M (depending on x) s.t. $|f_n(x) - f(x)| < \varepsilon/2$ whenever $n \geq M$.

Let $R = \max(N, M)$, then for $n \geq N$,

$$\begin{aligned} |f_n(x) - f(x)| &\leq |f_n(x) - f_R(x)| + |f_R(x) - f(x)| \\ &< \varepsilon/2 + \varepsilon/2 = \varepsilon \end{aligned}$$

for $n, R \geq N$. i.e. $|f_n(x) - f(x)| < \varepsilon$ for all $x \in [0, 1]$ i.e. $\|f_n - f\|_\infty < \varepsilon$.
So $(f_n) \rightarrow f$ uniformly.

$f_n \in C[0, 1] \implies f \in C[0, 1]$. So $(f_n) \rightarrow f \in C[0, 1]$ uniformly. \square

3.3 Uniform continuity

Suppose V, W are normed spaces, $A \subset V$.

Definition. $f : A \rightarrow W$ is *uniformly continuous* if for every $\varepsilon > 0$, $\exists \delta > 0$ s.t. $\|f(\mathbf{v}) - f(\mathbf{v}')\| < \varepsilon$ whenever $\|\mathbf{v} - \mathbf{v}'\| < \delta$.

Example. Let $f : \mathbb{R} \rightarrow \mathbb{R}$ by $f(x) = x^2$. Then $f(x + \delta) - f(x) = 2x\delta + \delta^2$. For fixed δ , $2x\delta + \delta^2 \rightarrow \infty$ as $x \rightarrow \infty$. So $f(x) = x^2$ is not uniformly continuous.

Example. Let $f : (0, 1] \rightarrow \mathbb{R}$ with $f(x) = \frac{1}{x}$. This is not uniformly continuous as well (consider $x \rightarrow 0$).

Theorem. If C is s.compact, and $f : C \rightarrow W$ is continuous, then f is uniformly continuous.

Proof. Suppose f is not uniformly continuous. Then there exists $\varepsilon > 0$ s.t. for all $n > 0$ we can find $\mathbf{v}_n, \mathbf{w}_n \in C$ with $\|\mathbf{v}_n - \mathbf{w}_n\| < \frac{1}{n}$, and $\|f(\mathbf{v}_n) - f(\mathbf{w}_n)\| \geq \varepsilon$ (else f is uniformly continuous).

Since C is s.compact, (\mathbf{v}_n) has a convergent subsequence $(\mathbf{v}_{n_i}) \rightarrow \mathbf{v}^* \in C$.

f is continuous and $\mathbf{v}^* \in C$, so $\exists \delta > 0$ s.t. $\|f(\mathbf{v}) - f(\mathbf{v}^*)\| < \varepsilon/2$ whenever $\mathbf{v} \in B_\delta(\mathbf{v}^*)$.

If $\mathbf{v}, \mathbf{v}' \in B_\delta(\mathbf{v}^*)$, then

$$\begin{aligned} \|f(\mathbf{v}) - f(\mathbf{v}')\| &\leq \|f(\mathbf{v}) - f(\mathbf{v}^*)\| + \|f(\mathbf{v}^*) - f(\mathbf{v}')\| \\ &< \varepsilon/2 + \varepsilon/2 = \varepsilon \end{aligned}$$

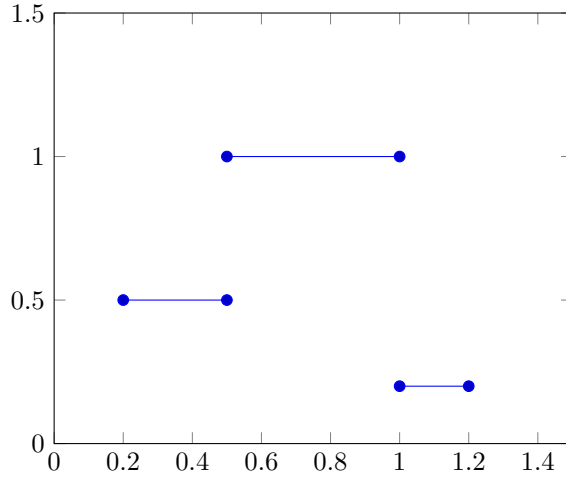
$(\mathbf{v}_{n_i}) \rightarrow \mathbf{v}^*$, so pick I_1 s.t. $\|\mathbf{v}_{n_i} - \mathbf{v}^*\| < \delta/2$ when $i \geq I_1$.

Pick I_2 s.t. $1/I_2 < \delta/2$. Then for $i \geq \max(I_1, I_2)$, we have $\|\mathbf{v}_{n_i} - \mathbf{v}^*\| < \delta/2$ and $\|\mathbf{v}_{n_i} - \mathbf{w}_{n_i}\| < \frac{1}{n_i} < \frac{1}{i} < \frac{1}{I_2} < \frac{\delta}{2}$.

So $\|\mathbf{w}_{n_i} - \mathbf{v}^*\| < \|\mathbf{w}_{n_i} - \mathbf{v}_{n_i}\| + \|\mathbf{v}_{n_i} - \mathbf{v}^*\| < \delta/2 + \delta/2 = \delta$, i.e. $\mathbf{w}_{n_i}, \mathbf{v}_{n_i} \in B_\delta(\mathbf{v}^*)$, $\|f(\mathbf{v}_{n_i}) - f(\mathbf{w}_{n_i})\| \geq \varepsilon$. Contradiction. So f must be uniformly continuous. \square

3.4 Application: Integration

Recall from Analysis I: We say $f : [a, b] \rightarrow \mathbb{R}$ is piecewise constant if $\exists a = a_0 < a_1 < \dots < a_n = b$ and $c_1, \dots, c_n \in \mathbb{R}$ s.t. $f(x) = c_i$ if $x \in (a_{i-1}, a_i)$.



Let $P[a, b] = \{f : [a, b] \rightarrow \mathbb{R} \mid f \text{ is piecewise constant}\}$. If $f \in P[a, b]$ is as above, then

$$I(f) = \sum_{i=1}^n c_i (a_i - a_{i-1}) = \int f$$

Lemma. If $f, g \in P[a, b]$, $\lambda \in \mathbb{R}$, then

$$f - \lambda g \in P[a, b]$$

and $I(f - \lambda g) = I(f) - \lambda I(g)$.

Write $f \geq g$ if $f(x) \geq g(x)$ for all $x \in [a, b]$.

Lemma. If $f \geq 0$, $I(f) \geq 0$.

So if $f, g \in P[a, b]$, $f \geq g$, then $I(f) \geq I(g)$.

Definition. (Riemann Integral) Suppose $f : [a, b] \rightarrow \mathbb{R}$ is bounded. Let

$$\begin{aligned} \mathcal{U}(f) &= \{g \in P[a, b] \mid g \geq f\}, \\ \mathcal{L}(f) &= \{g \in P[a, b] \mid g \leq f\} \end{aligned}$$

since f is bounded, these are not empty.

Let

$$\begin{aligned} U(f) &= \{I(g) \mid g \in \mathcal{U}(f)\}, \\ L(f) &= \{I(g) \mid g \in \mathcal{L}(f)\} \end{aligned}$$

If $g^+ \in \mathcal{U}(f)$ and $g^- \in \mathcal{L}(f)$, then $g^+ \geq f \geq g^-$. So $I(g^+) \geq I(g^-)$. If $u \in U(f)$ and $l \in L(f)$, then $u \geq l$. So $U(f)$ is bounded below, $L(f)$ is bounded above.

Now let

$$\begin{aligned} u(f) &= \inf U(f), \\ l(f) &= \sup L(f) \end{aligned}$$

Note that $u(f) \geq l(f)$.

We say f is Riemann integrable if $u(f) = l(f)$, in which case we define

$$\int_a^b f(x) dx = u(f) = l(f)$$

If $f \in P[a, b]$, then $u(f) = I(f) = l(f)$, so f is RI.

Theorem. If $f \in C[a, b]$, then f is RI.

Lemma. Given $\varepsilon > 0$, $\exists g^+ \in \mathcal{U}(f)$ and $g^- \in \mathcal{L}(f)$ s.t. $I(g^+) - I(g^-) < \varepsilon$.

Proof. $[a, b]$ is closed and bounded in \mathbb{R} , so it is s.compact. By last lecture's theorem, $f : [a, b] \rightarrow \mathbb{R}$ is uniformly continuous.

So pick δ s.t.

$$|f(x) - f(y)| < \frac{\varepsilon}{b-a}$$

whenever $|x - y| < \delta$. Choose $a = a_0 < a_1 < \dots < a_n = b$ such that $a_{i+1} - a_i < \delta$ for all i .

Define

$$c_i^+ = \max_{x \in [a_{i-1}, a_i]} f(x),$$

$$c_i^- = \min_{x \in [a_{i-1}, a_i]} f(x)$$

(These exist by Maximum value theorem) So

$$c_i^+ = f(x^+) \geq f(x^-) \forall x \in [a_{i-1}, a_i],$$

$$c_i^- = f(x^-) \leq f(x) \forall x \in [a_{i-1}, a_i]$$

$$x^+, x^- \in [a_{i-1}, a_i] \implies |x^+ - x^-| < \delta.$$

Define

$$g^+(x) = c_i^+ \text{ if } x \in [a_{i-1}, a_i],$$

$$g^-(x) = c_i^- \text{ if } x \in [a_{i-1}, a_i]$$

Then $|x^+ - x^-| < \delta \implies c_i^+ - c_i^- < \frac{\varepsilon}{b-a}$ for all i . So to sum up, $g^+ \geq f \geq g^-$ and $g^+ - g^- \leq \frac{\varepsilon}{b-a}$.

Thus $g^+ \in \mathcal{U}(f)$, $g^- \in \mathcal{L}(f)$, and

$$I(g^+) - I(g^-) = I(g^+ - g^-) \leq I\left(\frac{\varepsilon}{b-a}\right) = \varepsilon$$

□

Now prove the theorem:

Proof. $I(g^+) \geq u(f) \geq l(f) \geq I(g^-)$. So $u(f) - l(f) \leq I(g^+) - I(g^-) < \varepsilon$ for all $\varepsilon > 0$, which implies $u(f) = l(f)$. □

Corollary. If $f \in C[a, b]$, $\exists f_k \in P(a, b)$ s.t. $(f_k) \rightarrow f$ uniformly on $[a, b]$.

Proof. For each k , choose g_k^+ as in the proof of lemma with $\varepsilon = \frac{1}{k}$. Then $(g_k^+) \rightarrow f$ uniformly. \square

Example. (Speed and Distance) Suppose $f[a, b] \rightarrow \mathbb{R}^n$ is continuous. $f(t) = (f_1(t), \dots, f_n(t))$ where all f_i are continuous.

Define $\int_a^b f(t) dt = \left(\int_a^b f_1(t) dt, \dots, \int_a^b f_n(t) dt \right)$ (Integrating pointwise).

If $f(t) = \mathbf{v}(t)$ = velocity of a particle in \mathbb{R}^n at time t , then $\mathbf{p}(b) - \mathbf{p}(a) = \int_a^b \mathbf{v}(t) dt$ is the displacement of particle from its position at $t = a$. $\|\mathbf{v}(t)\|$ is the speed of particle.

Proposition. If $f : [a, b] \rightarrow \mathbb{R}^n$ is continuous, then

$$\left\| \int_a^b f(t) dt \right\| \leq \int_a^b \|f(t)\| dt$$

Lemma. If $x_i, y_i \in \mathbb{R}$ satisfy:

(1) $x_i \leq y_i$ for all i ;

(2) $(x_i) \rightarrow x$ and $(y_i) \rightarrow y$

Then $x \leq y$.

Proof. $y_i - x_i \geq 0, (y_i - x_i) \rightarrow y - x \implies y - x \geq 0$. \square

Lemma. The proposition holds if f is piecewise constant (maybe not continuous).

Proof. Suppose $f(t) = \mathbf{v}_i$ for $t \in (a_{i-1}, a_i)$. Then

$$\begin{aligned} \left\| \int_a^b f(t) dt \right\| &= \|I(f)\| \\ &= \left\| \sum_{i=1}^n (a_{i+1} - a_i) \mathbf{v}_i \right\| \\ &\leq \sum_{i=1}^n (a_i - a_{i-1}) \|\mathbf{v}_i\| \\ &= I(\|f\|) \\ &= \int_a^b \|f\| dt. \end{aligned}$$

\square

Proof of proposition:

Proof. Choose a sequence of piecewise constant functions $f_k : [a, b] \rightarrow \mathbb{R}^n$ s.t. $(f_k) \rightarrow f$ uniformly.

Then

$$\int_a^b f_k \rightarrow \int_a^b f$$

(uniformly convergence $\implies L^1$ convergence) and

$$\left(\left\| \int_a^b f_k \right\| \right) \rightarrow \left(\left\| \int_a^b f \right\| \right)$$

since $\|\cdot\|$ is continuous.

Also $(\|f_k\|) \rightarrow \|f\|$ uniformly ($\|\cdot\|$ is continuous). So

$$\left(\int_a^b \|f_k\| \right) \rightarrow \int_a^b \|f\|$$

So now take $x_k = \int_a^b f_k$, $x = \int_a^b f$, $y_k = \int_a^b \|f_k\|$, $y = \int_a^b \|f\|$.

Then $x_k \leq y_k$, so $x \leq y$. □

4 Differentiation

Slogan: *The derivative is a linear map.*

4.1 Derivative

Definition. Let $U \subset \mathbb{R}^n$ be open, $f : U - \{x_0\} \rightarrow \mathbb{R}^m$. We say

$$\lim_{x \rightarrow x_0} f(x) = y$$

if the function $\bar{f} : U \rightarrow \mathbb{R}^m$ given by

$$\bar{f}(x) = \begin{cases} f(x) & x \neq x_0 \\ y & x = x_0 \end{cases}$$

is continuous at x_0 .

Note that we don't care which norms on \mathbb{R}^n or \mathbb{R}^m we use: all the norms on \mathbb{R}^n are Lipschitz equivalent, so they determine the same continuous functions.

Definition. Suppose $U \subset \mathbb{R}^n$ is open, $x_0 \in U$ and $f : U \rightarrow \mathbb{R}^m$. We say f is differentiable at x_0 if there is a linear map $L : \mathbb{R}^n \rightarrow \mathbb{R}^m$ s.t.

$$\lim_{v \rightarrow 0} \frac{f(x_0 + v) - (f(x_0) + L(v))}{\|v\|} = 0$$

If such an L exists, it is unique.

Proof. Suppose L_1, L_2 exist. Subtracting the two limit equations gives

$$\lim_{v \rightarrow 0} \frac{L_2(v) - L_1(v)}{\|v\|} = 0$$

If $v \in \mathbb{R}^n$, $v \neq 0$, then $tv \rightarrow 0$ as $t \rightarrow 0^+$. So

$$\lim_{t \rightarrow 0^+} \frac{L_2(tv) - L_1(tv)}{\|tv\|} = 0$$

Since L_1, L_2 are linear maps, simplify that and we get $L_2(v) = L_1(v)$. But v is arbitrary. So $L_1 = L_2$. \square

When the equation in the definition of differentiability holds, we say

$$Df|_{x_0} = L$$

is the *derivative* of f at x_0 . Note that $Df|_{x_0}$ is a *linear map* from \mathbb{R}^n to \mathbb{R}^m .

Equivalently, f is differentiable at x_0 with $Df|_{x_0} = L$ if

$$f(x_0 + v) = f(x_0) + L(v) + \|v\|\alpha(v)$$

where $\lim_{v \rightarrow 0} \alpha(v) = 0$.

Proposition. Suppose $f : U \rightarrow \mathbb{R}^m$ is differentiable at $x_0 \in U$. Then f is continuous at x_0 .

Lemma. Suppose $L : \mathbb{R}^n \rightarrow (W, \|\cdot\|)$ is a linear map where W is a normed space. Then $\lim_{v \rightarrow 0} L(v) = 0$.

Note that the lemma is false if \mathbb{R}^n is replaced by an arbitrary normed space.

Proof. Let $v = (v_1, \dots, v_n) = \sum_{i=1}^n v_i e_i$. Then

$$\begin{aligned} \|L(v)\| &= \left\| \sum_{i=1}^n v_i L(e_i) \right\| \\ &\leq \sum_{i=1}^n |v_i| \cdot \|L(e_i)\| \\ &\leq C \sum_{i=1}^n |v_i| \\ &= C \|v\|_1 \end{aligned}$$

Where $C = \max\{\|L(e_1)\|, \dots, \|L(e_n)\|\}$.

Given $\varepsilon > 0$, pick $\delta > \varepsilon/C$. If $\|v\|_1 < \delta$ then $\|L(v)\| < \varepsilon$, so $\lim_{v \rightarrow 0} L(v) = 0$. \square

Prove of proposition:

Proof. Since f is differentiable at x_0 , we have

$$f(x_0 + v) = f(x_0) + L(v) + \|v\|\alpha(v)$$

where $\lim_{v \rightarrow 0} \alpha(v) = 0$. Now take the limit $v \rightarrow 0$ of both sides we have

$$\lim_{v \rightarrow 0} f(x_0 + v) = f(x_0)$$

So f is continuous at x_0 . \square

4.2 The derivative as a matrix

Suppose $U \subset \mathbb{R}^n$ is open, $f : U \rightarrow \mathbb{R}^m$.

We say f is differentiable if f is differentiable at all $x \in U$.

If so, we have $Df : U \rightarrow \mathcal{L}(\mathbb{R}^n, \mathbb{R}^m)$.

From Linear Algebra we know that there is a bijection between $\mathcal{L}(\mathbb{R}^n, \mathbb{R}^m)$ and the set of $m \times n$ real matrix:

$$[a_{ij}] \longleftrightarrow L(e_j) = \sum a_{ij} e_i$$

Now let's consider given $f : U \rightarrow \mathbb{R}^m$, how we compute $Df = [a_{ij}(x)]$. We first reduce to the case $m = 1$ by writing

$$f(x) = (f_1(x), \dots, f_n(x))$$

Then think about $F : U \rightarrow \mathbb{R}$.

Proposition. f is differentiable at x_0 if and only if f_i is differentiable for all $1 \leq i \leq m$. If so,

$$Df|_{x_0} = (Df_1|_{x_0}, \dots, Df_m|_{x_0}).$$

Proof. Suppose $g : U \rightarrow \mathbb{R}^m$. Using the uniform norm on \mathbb{R}^m , we see that $\lim_{v \rightarrow 0} g(v) = 0$ iff $\lim_{v \rightarrow 0} g_i(v) = 0$ for all $1 \leq i \leq m$.

Now let $L : \mathbb{R}^n \rightarrow \mathbb{R}^m$. Then

$$\lim_{v \rightarrow 0} \frac{f(x_0 + v) - (f(x_0) + L(v))}{\|v\|} = 0$$

if and only if

$$\lim_{v \rightarrow 0} \frac{f_i(x_0 + v) - (f_i(x_0) + L_i(v))}{\|v\|} = 0$$

for all $1 \leq i \leq m$, i.e. f_i is differentiable at x_0 and $Df_i|_{x_0} = L_i$. \square

Summary:

$$Df|_{x_0} = \begin{bmatrix} Df_1|_{x_0} \\ \dots \\ Df_m|_{x_0} \end{bmatrix}$$

where $Df_i|_{x_0} : \mathbb{R}^n \rightarrow \mathbb{R}$ is a $1 \times n$ matrix $[a_1, \dots, a_n]$.

Definition. (Directional Derivative)

Suppose $F : U \rightarrow \mathbb{R}$. If $v \in \mathbb{R}^n$, the *directional derivative* of F in direction v at x is

$$\begin{aligned} D_v F|_x &= \lim_{t \rightarrow 0} \frac{F(x + tv) - F(x)}{t} \\ &= \frac{d}{dt}(F(x + tv))|_{t=0} \end{aligned}$$

$D_v F$ measures the rate of change of F if I walk away from x at velocity v .

It's also helpful to consider

$$D_v^+ F = \lim_{t \rightarrow 0^+} \frac{F(x + tv) - F(x)}{t}$$

and similarly for $D_v^- F$. We can prove that

$$D_v^- F|_x = -D_{-v}^+ F|_x$$

Note: $D_v F$ exists iff $D_v^+ F, D_v^- F$ both exist and are equal.

Example. Consider a special case $v = e_i$. Then

$$\begin{aligned} D_i F|_x &= \frac{\partial F}{\partial x_i}|_x \\ &= D_{e_i} F|_x \\ &= \frac{d}{dt}(F(x_1, \dots, x_i + t, \dots, x_n))|_{t=0} \\ &= \frac{d}{dt}(F(x_1, \dots, x_{i-1}, t, x_{i+1}, \dots, x_n))|_{t=x} \end{aligned}$$

is the i th partial derivative of F .

Proposition. If $F : U \rightarrow \mathbb{R}$ is differentiable at x , then $D_v F|_x = DF|_x(v)$.

Proof. If $v = 0$ then both sides are 0.

If $v \neq 0$, then $tv \rightarrow 0$ as $t \rightarrow 0^+$, so differentiability of F implies

$$\lim_{t \rightarrow 0^+} \frac{F(x + tv) - (F(x) + L(tv))}{\|tv\|} = 0$$

where $L = DF|_x$. So

$$\lim_{t \rightarrow 0^+} \frac{F(x + tv) - F(x)}{t} - L(v) = 0$$

i.e. $D_v^+ F|_x = DF|_x(v)$. Then $D_v^- F|_x = -D_{-v}^+ F|_x = -L(-v) = L(v)$. \square

If $DF|_x = [a_1, \dots, a_n]$ then $a_i = DF|_x(e_i) = D_{e_i} F|_x = D_i F|_x$. So we have

$$DF|_x = [D_1 F|_x, \dots, D_n F|_x]$$

Summary: if $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$, then

$$Df = \begin{bmatrix} Df_1 \\ \dots \\ Df_m \end{bmatrix} = [D_j f_i]$$

Example. Let $f : \mathbb{R}^3 \rightarrow \mathbb{R}^2$ with $f(x, y, z) = (x^2 + y^2 + z^2, xyz)$. Then

$$Df = \begin{bmatrix} 1 & 2y & 3z^2 \\ yz & xz & xy \end{bmatrix}$$

Note: Just because $D_i F|_x$ all exists doesn't mean that F is differentiable at x .

Example. Let $F : \mathbb{R}^2 \rightarrow \mathbb{R}$ be given by

$$F(x, y) = \begin{cases} 0 & xy = 0 \\ H(x, y) & \text{otherwise} \end{cases}$$

where $H(x, y)$ is any arbitrary horrible function. Then

$$D_1 F|_0 = D_2 F|_0 = 0$$

but F may not even be continuous.

We can even have $D_v F$ well defined for every v , but F is not differentiable.

Example. Let $S^1 = \{v \in \mathbb{R}^2 \mid \|v\| = 1\}$. Choose $h : S \rightarrow \mathbb{R}$ to be any function. Define $F : \mathbb{R}^2 \rightarrow \mathbb{R}$ by

$$F(v) = \begin{cases} \|v\| h(\frac{v}{\|v\|}) & v \neq 0 \\ 0 & v = 0 \end{cases}$$

Then $D_v^+ F|_0 = \|v\| h(\frac{v}{\|v\|})$. If we let $h(-v) = -h(v)$ then $D_v^+ F = D_v^- F$, so $D_v F$ is well defined. Now if F is differentiable, $D_v F|_0 = DF|_0(v)$, so $h(v)$ would have to be a linear function on S^1 ; but h is arbitrary except the one condition above.

A criterion for differentiability: Let $U \subset \mathbb{R}^n$ be open.

Definition. $C^1(U) = \{f : U \rightarrow \mathbb{R} \mid \text{for } 1 \leq i \leq n, \text{ the partial derivative } D_i f|_x \text{ exists for all } x \in U \text{ and is a continuous function of } x\}$.

Example.

$$F(x, y, z) = e^{\cos x^2 y + z} - y^2 z \in C^1(\mathbb{R}^3)$$

Theorem. If $F \in C^1(U)$, then F is differentiable on U . Tools used in proof:

- Alternative characterisation of differentiability in 4.1;
- If $\lim_{v \rightarrow 0} g(v) = w_0$ and $\lim_{v \rightarrow w_0} f(v) = z$, then $\lim_{v \rightarrow 0} f(g(v)) = z$;
- Suppose $b : U \rightarrow \mathbb{R}$ is bounded on $B_r(v_0)$ for some $r > 0$. Then $\lim_{v \rightarrow v_0} b(v)\alpha(v) = 0$ if $\lim_{v \rightarrow v_0} \alpha(v) = 0$.

Proof. (of the bullet point):

Since b is bounded, there exists $M \in \mathbb{R}$ s.t. $|b(x)| \leq M$ for all $v \in B_r(v_0)$. Since $\lim_{v \rightarrow 0} \alpha(v) = 0$, given $\varepsilon > 0$, there exists $\delta > 0$ s.t. $\|\alpha(v)\| < \frac{\varepsilon}{M}$ whenever $v \in B_\delta(v_0)$. Then let $\delta' = \min(\delta, r)$. We have $\|b(v)\alpha(v)\| = \|b(v)\| \|\alpha(v)\| < \varepsilon$ for $v \in B_{\delta'}(v_0)$. So $\lim_{v \rightarrow v_0} b(v)\alpha(v) = 0$. \square

Proof. (for $n = 2$)

We want to estimate $F(x+v) - F(x)$ for small v . Since U is open, $\exists r > 0$ s.t. $B_r(x) \subset U$.

From now on we assume $\|v\| < r$ (since v is small that's reasonable). So $x' \in U$. Since $D_1 F$ exists, we write

$$F(x') - F(x) = F(x_1 + v_1, x_2) - F(x_1, x_2) = v_1 D_1 F|_x + |v_1| \alpha_1(v_1)$$

where $\lim_{v_1 \rightarrow 0} \alpha_1(v_1) = 0$. Similarly

$$F(x+v) - F(x') = v_2 \cdot D_2 F|_x + |v_2| \alpha_2(v_2)$$

where $\lim_{v_2 \rightarrow 0} \alpha_2(v_2) = 0$.

Mistake! Here $\alpha_2(v_2)$ depends on v_1 .

Instead, apply 1-variable mean value theorem to $f(t) = F(x_1 + v_1, x_2 + t)$ to write

$$F(x+v) - F(x') = v_2 D_2 F|_{x''(v)}$$

where $x''(v) = (x_1 + v_1, x_2 + h(v))$ where $0 < h(v) < v_2$. Then as before, we can add to get

$$F(x + v) - F(x) = L(v) + \|v\|E(v)$$

where

$$\begin{aligned} E(v) &= \frac{|v_1|}{\|v\|} \alpha_1(v_1) + \frac{|v_2|}{\|v\|} (D_2F|_{x'(v)} - D_2F|_x) \\ &= E_2(v) + E_1(v) \end{aligned}$$

Note that $\|x''(v) - x\|_2 = (v_1^2 + h(v)^2)^{0.5} \leq \|v\|_2$. So $\lim_{v \rightarrow 0} x''(v) = x$.

Now D_2F is continuous, so $\lim_{v \rightarrow 0} D_2F|_{x''} - D_2F|_x = 0$.

We'll show that as $v \rightarrow 0$, $E_1(v), E_2(v) \rightarrow 0$, then we are done.

- E_1 : As $v \rightarrow 0$, $x' \rightarrow x$. Now D_2F is continuous, so

$$\lim_{x' \rightarrow x} (D_2F|_{x'} - D_2F|_x) = 0$$

So

$$\lim_{v \rightarrow 0} (D_2F|_{x'} - D_2F|_x) = 0$$

Now $\frac{|v_2|}{\|v\|} < 1$ for all $v \in \mathbb{R}^2 \setminus \{0\}$, so by lemma $E_1 \rightarrow 0$.

- E_2 : $\lim_{v \rightarrow 0} v_1 = 0$ and $\lim_{v_1 \rightarrow 0} \alpha(v_1) = 0$, so $\lim_{v \rightarrow 0} \alpha_1(v_1) = 0$. Same as above we get $E_2 \rightarrow 0$.

(Refer to DC notes last page of Section 6.1 (p66).)

□

Example. Let $V = M_{n \times n}(\mathbb{R}) \cong \mathbb{R}^{n^2}$, $f : V \rightarrow V$ by $f(x) = x^2$. Then

$$f(x + v) = (x + v)^2 = x^2 + xv + vx + v^2 = f(x) + L_x(v) + v^2$$

where

$$L_x(v) = xv + vx$$

is linear in V . Compare with the definition we get

$$DF|_x = Lx.$$

4.3 The Chain Rule

Theorem. (Chain Rule)

Suppose $g : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is differentiable at x , and $f : \mathbb{R}^m \rightarrow \mathbb{R}^l$ is differentiable at $g(x)$. Then $f \circ g : \mathbb{R}^n \rightarrow \mathbb{R}^l$ is differentiable at x , and

$$D(f \circ g)|_x = Df|_{g(x)} \circ Dg|_x$$

Example. Suppose $r : \mathbb{R} \rightarrow \mathbb{R}^n$ by $r(t) = (r_1(t), \dots, r_n(t))$, $F : \mathbb{R}^n \rightarrow \mathbb{R}$, $F \circ r : \mathbb{R} \rightarrow \mathbb{R}$.

Then $D(F \circ r)|_t$ is a linear map $\mathbb{R} \rightarrow \mathbb{R}$ given by 1×1 matrix $[\frac{d}{dt}(F \circ r)]$, $Dr|_t : \mathbb{R} \rightarrow \mathbb{R}^n$ is given by

$$\begin{bmatrix} r'_1|_t \\ \dots \\ r'_n|_t \end{bmatrix}$$

and $DF|_t : \mathbb{R}^n \rightarrow \mathbb{R}$ given by

$$[D_1F|_{r(t)}, \dots, D_nF|_{r(t)}]$$

So $D(F \circ r)$ is given by matrix multiplication:

$$\begin{aligned} D(F \circ r) &= \sum_{i=1}^n D_iF|_{r(t)} \cdot r'_i(t)|_t \\ &= \sum \frac{\partial F}{\partial x_i} r'_i \end{aligned}$$

Now back to the theorem. Since g is differentiable, $g(x+v) = g(x) + (L_1(v) + \|v\|\alpha(v)) (= e(v))$ at x where $L_1 = Dg|_x : \mathbb{R}^n \rightarrow \mathbb{R}^m$ and $\lim_{v \rightarrow 0} \alpha(v) = 0$.

Lemma. $\lim_{v \rightarrow 0} e(v) = 0$.

Proof. g is differentiable at $x \implies g$ is continuous at x . Done. \square

Lemma. $\exists r > 0$, s.t. $\frac{\|e(v)\|}{\|v\|}$ is bounded on $B_r(0)$.

Proof.

$$\begin{aligned} \frac{\|e(v)\|}{\|v\|} &= \|L_1(\frac{v}{\|v\|}) + \alpha(v)\| \\ &\leq \|L_1(\frac{v}{\|v\|})\| + \|\alpha(v)\| \end{aligned}$$

write $v' = \frac{v}{\|v\|}$, so $\|v'\| = 1$.

L_1 is linear, so continuous. $\{v \in \mathbb{R}^n \mid \|v\| = 1\}$ is closed and bounded in \mathbb{R}^n , so by MVT, $\exists M$ s.t. $\|L_1(v')\| \leq M$ for all v' with $\|v'\| = 1$.

$\lim_{v \rightarrow 0} \alpha(v) = 0$, so $\exists r$ s.t. $\|\alpha(v)\| < 1$ for $v \in B_r(0)$.

Then for $v \in B_r(0)$, $\frac{\|e(v)\|}{\|v\|} \leq M + 1$. \square

Proof. (of Chain Rule)

f is differentiable at $g(x)$, so

$$f(g(x) + w) = f(g(x)) + L_2(w) + \|w\|B(w)$$

where $L_2 = Df|_{g(x)}$ and $\lim_{w \rightarrow 0} B(w) = 0$.

$$\begin{aligned}
f(g(x+v)) &= f(g(x) + e(v)) \\
&= fg(x) + L_2(e(v)) + \|e(v)\|B(e(v)) \\
&= fg(x) + L_2(L_1(v)) + L_2(\|v\|\alpha(v)) + \|e(v)\|B(e(v)) \\
&= fg(x) + (Df|_{g(x)} \cdot Dg|_x)(v) + \|v\|E(v)
\end{aligned}$$

where

$$E(v) = L_2(\alpha(v)) + \frac{\|e(v)\|}{\|v\|} B(e(v))$$

we must show that $\lim_{v \rightarrow 0} E(v) = 0$ and then we are done.

We know $\lim_{v \rightarrow 0} \alpha(v) = 0$. L_2 is linear, hence continuous, so $\lim_{w \rightarrow 0} L_2(w) = L_2(0) = 0$. Thus $\lim_{v \rightarrow 0} L_2(\alpha(v)) = 0$.

By the above second lemma, $\exists r > 0$ s.t. $\frac{\|e(v)\|}{\|v\|}$ is bounded on $B_r(0)$. By the above first lemma $\lim_{v \rightarrow 0} e(v) = 0$.

We know $\lim_{w \rightarrow 0} B(w) = 0 \implies \lim_{v \rightarrow 0} B(e(v)) = 0$.

Then by last lecture's lemma,

$$\lim_{v \rightarrow 0} \frac{\|e(v)\|}{\|v\|} B(e(v)) = 0$$

So $\lim_{v \rightarrow 0} E(v) = 0$. □

Application of Chain Rule:

- The gradient.

Suppose $F : U \rightarrow \mathbb{R}$, where $U \subset \mathbb{R}^n$ is open. $DF|_x \in \mathcal{L}(\mathbb{R}^n, \mathbb{R})$.

Recall from LA that $\mathbb{R}^n \cong \mathcal{L}(\mathbb{R}^n, \mathbb{R})$ by $v \rightarrow \phi_v : \phi_v(w) = v \cdot w$. That sends $\nabla F|_x$ to $DF|_x = [D_1 F|_x, \dots, D_n F|_x]$ where $\nabla F|_x = (D_1 F|_x, \dots, D_n F|_x)$ is the *gradient* of F at x .

So

$$D_v F|_x = DF|_x(v) = \nabla F|_x \cdot v$$

- Mean value inequality.

Definition. (Convex)

lue

Proposition. Suppose $U \subset \mathbb{R}^n$ is open and convex, and $F : U \rightarrow \mathbb{R}$ is differentiable. If $\|\nabla F|_x\|_2 \leq M \forall x \in U_1$. Then

$$|F(x_1) - F(x_0)| \leq M \|x_1 - x_0\|_2$$

for all $x_0, x_1 \in U$.

Proof. Let $\gamma : [0, 1] \rightarrow \mathbb{R}^n$ be given by

$$\gamma(t) = (1-t)x_0 + tx_1$$

then γ is differentiable and $\gamma'(t) = x_1 - x_0$.

Let $f(t) = F(\gamma(t))$. By Chain rule, f is differentiable and $f'(t) = \nabla F|_{\gamma(t)} \cdot \gamma'(t)$.

By Cauchy-Schwartz,

$$\begin{aligned} |f'(t)| &\leq \|\nabla F|_{\gamma(t)}\| \cdot \|x_1 - x_0\| \\ &\leq M\|x_1 - x_0\| \end{aligned}$$

Apply 1-variable MVT to $f(t)$, we see that

$$|F(x_1) - F(x_0)| = |f(1) - f(0)| = |f'(c)|$$

for some $c \in [0, 1]$

$$\leq M\|x_1 - x_0\|$$

□

Corollary. If $U \subset \mathbb{R}^n$ is open and convex, $F : U \rightarrow \mathbb{R}$ has $D_i F \equiv 0$ for $1 \leq i \leq n$. Then $F(x) \equiv c$ for some $c \in \mathbb{R}$.

Proof. $D_i F \equiv 0 \implies F$ differentiable \implies

$$|F(x_1) - F(x_0)| \leq 0 \cdot \|x_1 - x_0\| = 0$$

for all $x_1, x_0 \in U$.

□

Remark. The hypothesis that U is convex is needed for the proposition, but can be weakened for the corollary.

Example. Suppose any 2 points x_1, x_0 in U can be joined by a differentiable path $\gamma : [0, 1] \rightarrow U$ with $\gamma(0) = x_0, \gamma(1) = x_1$. Then the corollary still holds.

Proof. Consider $f(t) = F(\gamma(t))$. Then $f'(t) = DF|_{\gamma(t)}(\gamma'(t))$ by the chain rule. $D_i F \equiv 0 \implies DF \equiv 0 \implies f'(t) \equiv 0 \implies f(t)$ is constant. So $F(x_0) = f(0) = f(1) = F(x_1)$ for any x_0, x_1 in U . □

However, the corollary does not hold if U is disconnected. In fact it holds whenever $U \subset \mathbb{R}^n$ is open and connected.

4.4 Higher Derivatives

Q: If the derivative is a linear map, what is the 2nd derivative?

A: 2nd derivative is a symmetric bilinear form.

Suppose $U \subset \mathbb{R}^n$ is open, $f : U \rightarrow \mathbb{R}^m$ is differentiable.

Fix $v \in \mathbb{R}^n$ and define $g_v : U \rightarrow \mathbb{R}^m$ by

$$g_v(x) = Df|_x(v).$$

Definition. f is twice differentiable if all g_v are differentiable. If so, define $D^2f|_x(v, w) = D_{g_v}(w)$, i.e.

$$D^2f|_x : \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R}^m$$

Example. $V = M_{n \times n}(\mathbb{R}) \cong \mathbb{R}^{n^2}$. $f : V \rightarrow V$ is given by $f(x) = x^2$. Then from previous section we know

$$g_A(X) = DF|_X(A) = XA + AX$$

Differentiate $g_A(X)$, get

$$\begin{aligned} g_A(X+B) &= A(X+B) + (X+B)A \\ &= (AX + XA) + AB + BA \\ &= g_A(X) + L_A(B) \end{aligned}$$

where $L_A(B) = AB + BA$ is linear in B .

So $D_{g_A}|_X(B) = AB + BA = D^2f|_X(A, B)$.

Note: $D^2f_X(A, B) = D^2f|_X(B, A)$.

Lemma. Suppose $f : U \rightarrow \mathbb{R}^m$ is twice differentiable, let $B(v, w) = D^2f|_x(v, w)$. Then B is a bilinear form.

Proof.

$$\begin{aligned} g_{v_1+\lambda v_2}(x) &= Df|_x(v_1 + \lambda v_2) \\ &= Df|_x(v_1) + \lambda Df|_x(v_2) = g_{v_1}(x) + \lambda g_{v_2}(x) \end{aligned}$$

So differentiating we get linearity in the first argument. Similarly we can prove linearity in the second argument. \square

Suppose $F : U \rightarrow \mathbb{R}$ is differentiable. Then the partial derivatives $D_i F : U \rightarrow \mathbb{R}$ are all defined.

Notation. Write $D_{ij}F = D_i(D_jF)$ if it exists.

Definition. $C^2(U) = \{F : U \rightarrow \mathbb{R} \mid \text{all 1st and 2nd order partial derivatives of } F \text{ are defined and continuous}\}$.

Proposition. If $F \in C^2(U)$, then F is twice differentiable and

$$D^2F|_x(v, w) = \sum_{1 \leq i, j \leq n} v_i w_j D_{ji}F(x)$$

Proof. Let $G_i = D_i F$. Then all 1st order partial derivatives of G_i are defined and continuous so G_i is differentiable.

Then for $v \in \mathbb{R}^n$, $G_{v(x)} = DF|_x(v) = \sum_{1 \leq i \leq n} v_i D_i F|_x = \sum v_i G_i(x)$.

So for a fixed value of v , $G_v(x)$ is a linear combination of the G_i s. Since all of them are differentiable, G_v is differentiable. So F is twice differentiable, and $D^2F|_x(v, w) = DG_v|_x(w) = \sum_{1 \leq j \leq n} w_j D_j G_v|_x = \sum_{1 \leq j, i \leq n} v_i w_j D_{ji}F|_x$. \square

Now $D_j(G_v) = D_j(\sum_{i=1}^n v_i G_i) = \sum_{i=1}^n v_i D_j G_i = \sum_{i=1}^n v_i D_{ji} F$.

Equivalently, $D^2 F|_x(v, w) = W^t B v$ where $B = [D_{ij} F|_x]$ is the *Hessian* matrix of 2nd order partial derivatives.

Example. $F(x, y) = x^2 y^3$. Then

$$B = \begin{pmatrix} 2y^3 & 6xy^2 \\ 6xy^2 & 6x^2 y \end{pmatrix}$$

Recall that if $U \subset \mathbb{R}^n$ is open and $F \in C^2(U)$, then $D^2 F|_x : \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R}$ is bilinear and given by

$$D^2 F|_x(v, w) = \sum_{1 \leq i, j \leq n} v_i w_j D_{ji} F|_x = w^T H(x) v$$

where $H(x) = [D_{ji} F|_x]$ is the *Hessian matrix*.

Theorem. (symmetry of mixed partials)

Suppose $U \subset \mathbb{R}^2$ is open and $F \in C^2(U)$. Then $D_{12} F = D_{21} F$.

Note that it's not enough for the partial derivatives to be defined. They must be continuous or the theorem may fail (see example sheet).

Lemma.

$$D_{12} F|_{(x_0, y_0)} = \lim_{v \rightarrow 0} \frac{S(v)}{v^2}$$

where

$$S(v) = F(x_0 + v, y_0 + v) - F(x_0 + v, y_0) - F(x_0, y_0 + v) + F(x_0, y_0)$$

Proof. Since U is open, there exists $\varepsilon > 0$ s.t. $B_\varepsilon((x_0, y_0), \|\cdot\|_\infty) \subset U$.

From now on, assume $|v| < \varepsilon/2$.

Consider $A(y) = F(x_0 + v, y) - F(x_0, y)$. Fix v with $|v| < \varepsilon/2$. Then A is differentiable on $(y_0 - \varepsilon/2, y_0 + \varepsilon/2)$, and

$$A'(y) = D_2 F(x_0 + v, y) - D_2 F(x_0, y)$$

Note that $S(v) = A(y_0 + v) - A(y_0)$. So by MVT,

$$S(v) = v A'(y^*)$$

for some $y^* \in [y_0, y_0 + v]$

$$\begin{aligned} &= v [D_2 F(x_0 + v, y^*) - D_2 F(x_0, y^*)] \\ &= v [B(x_0 + v) - B(x_0)] \end{aligned}$$

where $B(x) = D_2 F(x, y^*)$.

B is differentiable on $(x_0 - \varepsilon/2, x_0 + \varepsilon/2)$, aget $B'(x) = D_{12}F(x, y^*)$. Applying MVT to B we get

$$S(v) = v^2 B'(x^*) = v^2 D_{12}F(x^*, y^*)$$

for some $x^* \in [x_0, x_0 + v]$. Note that we have

$$\|(x^*(v), y^*(v)) - (x_0, y_0)\|_\infty \leq \|v\|_\infty$$

So

$$\lim_{v \rightarrow 0} (x^*(v), y^*(v)) = (x_0, y_0)$$

Then

$$\begin{aligned} \lim_{v \rightarrow 0} \frac{S(v)}{v^2} &= \lim_{v \rightarrow 0} D_{12}F(x^*(v), y^*(v)) \\ &= D_{12}F(x_0, y_0) \end{aligned}$$

Since D_{12} is continuous. □

Proof. (of theorem)

The expression $S(v)$ is symmetric under interchanging roles of x and y . Similar arguments as in the above proof shows

$$D_{21}F(x_0, y_0) = \lim_{v \rightarrow 0} \frac{S(v)}{v^2} = D_{12}F(x_0, y_0)$$

So they are equal. □

Corollary. If $U \subset \mathbb{R}^n$ is open, $G \in C^2(U)$, then $D_{ij}G = D_{ji}G$ for all $1 \leq i, j \leq n$.

Proof. Apply the theorem to $F(z_1, z_2 = G(x_1, x_2, \dots, z_1(i), \dots, z_2(j), \dots, x_n))$. □

In other words, if $G \in C^2(U)$, the Hessian matrix $H = [D_{ji}G|_x]$ is symmetric: $H^T = H$.

Corollary. $D^2G|_x$ is symmetric. i.e. $D^2G|_x(v, w) = D^2G|_x(w, v)$.

Proof. $D^2G|_x(v, w) = w^T H v$ is a 1×1 matrix, so symmetric. Take its transpose and we get the other side of the equation. □

Higher derivatives are defined inductively: If $F : U \rightarrow \mathbb{R}$ is $(k-1)$ times differentiable, then

$$D^k F|_x(v_1, \dots, v_k) = DG|_x(v_k)$$

(if exists) where

$$G(x) = D^{k-1}F|_x(v_1, \dots, v_{k-1})$$

The same proof as for $k = 2$ shows that if $F \in C^k(U)$ then F is k times differentiable, and

$$D^k F|_x(v_1, \dots, v_k) = \sum_{\alpha \in \{1, \dots, n\}^k} v^\alpha D_\alpha F|_x$$

where

$$v^\alpha = \prod_{i=1}^k v_{i, \alpha_i}$$

where

$$\begin{aligned} v_i &= (v_{i,1}, \dots, v_{i,n}) \\ \alpha &= (\alpha_1, \dots, \alpha_k) \end{aligned}$$

If we let $A(v_1, \dots, v_k) = D^k F|_\alpha(v_1, \dots, v_k)$, then A is

1) Symmetric: $A(v_1, \dots, v_k) = A(v_{i_1}, \dots, v_{i_k})$ where (i_1, \dots, i_k) is any permutation of $(1, \dots, k)$;

2) Multilinear: $(v_1 + \lambda v'_1, v_2, \dots, v_k) = A(v_1, v_2, \dots, v_k) + \lambda A(v'_1, v_2, \dots, v_k)$.

Proposition. Suppose $F \in C^k(U)$ and define

$$f(t) = F(x_0 + tv)$$

for $x_0 \in U$. Since U is open, f is defined on $(-\varepsilon, \varepsilon)$.

Then f is k -times differentiable and

$$f^k(t) = D^k F|_{x_0+tv}(v, v, \dots, v) \text{ (k times)}$$

Proof. Recall that if $G \in C^1(U)$ and $g = G(x_0 + tv)$ then $g'(t) = D_v G|_{x_0+tv} = DG|_{x_0+tv}(v)$.

The proof is by induction on k :

$k = 1$ is exactly the above equation applied to $G = F$.

For the general case, suppose the proposition holds for $k - 1$. Then let

$$\begin{aligned} h(t) &= f^{(k-1)}(t) \\ &= D^{k-1} F|_{x_0+tv}(v, \dots, v) \\ &= H(x_0 + tv) \end{aligned}$$

where $H(x) = D^{k-1} F|_x(v, \dots, v)$.

Apply the above equation to $G = H$, get

$$f^k(t) = h'(t) = DH|_{x_0+tv}(v) = D^k F|_{x_0+tv}(v, \dots, v) \text{ (k times)}$$

□

Theorem. (Taylor's Theorem)

If $F \in C^k(\mathbb{R}^n)$, then

$$F(x_0 + v) = \sum_{i=0}^{k-1} \frac{1}{i!} D^i F|_{x_0}(v, \dots, v) + \frac{1}{k!} D^k F|_{x_0+tv}(v, \dots, v)$$

for some $t \in [0, 1]$.

Proof. Consider $f(t) = F(x_0 + tv)$ as above. Then by Taylor's theorem in 1 variable, we have

$$f(1) = \sum_{i=0}^{k-1} \frac{1}{i!} f^{(i)}(0) \cdot 1^i + \frac{1}{k!} f^{(k)}(t) 1^k$$

for some $t \in [0, 1]$, i.e.

$$F(x_0 + v) = \sum_{i=0}^{k-1} \frac{1}{i!} D^i F|_{x_0}(v, \dots, v) + \frac{1}{k!} D^k F|_{x_0+tv}(v, \dots, v)$$

□

Remark. $D^k F|_{x_0}(v, \dots, v)$ is a degree k polynomial in the coefficients of v s.t. all the k^{th} order partial derivatives agree with k^{th} order partial derivatives of F at x_0 .

5 Metric spaces

5.1 Basics

Definition. *lue*

Example. *lue*

Definition. (open and closed sets)
lue

5.2 Lipschitz Maps

Suppose (X, d_X) and (Y, d_Y) are metric spaces.

Definition. $f : X \rightarrow Y$ is k -Lipschitz ($k \in \mathbb{R}^+$) if $d_Y(f(x_1), f(x_2)) \leq k d_X(x_1, x_2)$ for all $x_1, x_2 \in X$.
 f is Lipschitz if it's k -Lipschitz for some $k \in \mathbb{R}^+$.

Proposition. f is Lipschitz implies that f is uniformly continuous.

Proof. Suppose f is k -Lipschitz. If $d(x_1, x_2) < \varepsilon/k$, then $d(f(x_1), f(x_2)) < \varepsilon$. \square

Proposition. Suppose $U \subset \mathbb{R}^n$ is open, $F \in C^1(U)$, and $k = \bar{B}_r(\mathbf{x}_0) \subset U$. Then $F|_k$ is Lipschitz.

Proof. F is C^1 , so the map

$$\begin{array}{ccc} U \subset \mathbb{R}^n & & \rightarrow \mathbb{R} \\ \mathbf{x} \mapsto \nabla F|_{\mathbf{x}} & \rightarrow & \|\nabla F|_{\mathbf{x}}\| \end{array}$$

is continuous.

$k = \bar{B}_r(\mathbf{x}_0)$ is a closed and bounded subset of \mathbb{R}^n . By the Maximum Value Theorem, $\exists M \in \mathbb{R}$ s.t. $\|\nabla F|_{\mathbf{x}}\| \leq M$ for all $\mathbf{x} \in k$. $k = \bar{B}_r(\mathbf{x}_0)$ is convex, so by Mean Value Inequality,

$$|F(\mathbf{x}_1) - F(\mathbf{x}_2)| \leq M \|\mathbf{x}_1 - \mathbf{x}_2\|_2$$

i.e.

$$d(F(\mathbf{x}_1), F(\mathbf{x}_2)) \leq M d(\mathbf{x}_1, \mathbf{x}_2)$$

for $\mathbf{x}_1, \mathbf{x}_2 \in k$. \square

Proposition. If $f : Y \rightarrow Z$ is k_1 -Lipschitz, $g : X \rightarrow Y$ is k_2 -Lipschitz, then $f \circ g : X \rightarrow Z$ is $k_1 k_2$ -Lipschitz.

Proof.

$$\begin{aligned} d_2(f(g(x_1)), f(g(x_2))) &\leq k_1 d_y(g(x_1), g(x_2)) \\ &\leq k_1 k_2 d_x(x_1, x_2) \end{aligned}$$

So The composition of Lipschitz maps is Lipschitz. \square

Proposition. If $\|\cdot\|$ and $\|\cdot\|'$ are two norms on a vector space V , then $\|\cdot\|$ is Lipschitz equivalent to $\|\cdot\|'$ if and only if both the identity maps from V equipped with one norm to the other norm are Lipschitz.

Definition. Suppose V, W are finite dimensional normed vector spaces. If $L \in \mathcal{L}(V, W)$, the operator norm

$$\|L\|_{op} = \sup_{\mathbf{v} \in V, \mathbf{v} \neq 0} \frac{\|L(\mathbf{v})\|_W}{\|\mathbf{v}\|_V} = \max_{\|\mathbf{v}\|=1} \|L(\mathbf{v})\|_W.$$

The maximum exists since S^1 is closed and bounded in $V = \mathbb{R}^n$.

Lemma. $\|\cdot\|_{op}$ is a norm on $\mathcal{L}(V, W)$.

Proof. Omitted. \square

Proposition. If $\|L_1\|_{op} = k$, then L_1 is k -Lipschitz.

Proof.

$$\begin{aligned} \|L(\mathbf{v}_1) - L(\mathbf{v}_2)\| &= \|L(\mathbf{v}_1 - \mathbf{v}_2)\| \\ &\leq k \|\mathbf{v}_1 - \mathbf{v}_2\| \end{aligned}$$

Since $\|L\|_{op} = k$. \square

5.3 Contraction maps

Suppose X is a metric space and $f : X \rightarrow X$.

Definition. $x \in X$ is a *fixed point* of f if $f(x) = x$.

Definition. $f^n = f \circ f \circ \dots \circ f$ (n times) : $X \rightarrow X$ it the composition of f with itself n times.

If f is k -Lipschitz, then f^n is k^n -Lipschitz.

Definition. $f : X \rightarrow X$ is a *contraction map* if f is k -Lipschitz for some $k < 1$.

Theorem. Suppose X is a complete metric space, $f : X \rightarrow X$ is a contraction map. Then f has a unique fixed point.

Proof. Suppose f is k -Lipschitz for some $k < 1$.

Lemma. If $x \in X$, then $d(x, f^n(x)) \leq \frac{1}{1-k} d(x, f(x))$ regardless of n .

Proof. f^n is k^n Lipschitz, so

$$\begin{aligned} d(f^n(x), f^{(n+1)}(x)) &= d(f^n(x), f^n(f(x))) \\ &\leq k^n d(x, f(x)) \end{aligned}$$

So

$$\begin{aligned} d(x, f^n(x)) &\leq d(x, f(x)) + \dots + d(f^{n-1}(x), f^n(x)) \\ &\leq d(x, f(x)) + kd(x, f(x)) + \dots + k^{n-1}d(x, f(x)) \\ &= \frac{1 - k^n}{1 - k} d(x, f(x)) \\ &\leq \frac{1}{1 - k} d(x, f(x)) \end{aligned}$$

□

Proof of Theorem:

Pick $x \in X$ and consider $(f^n(x))$.

This sequence is Cauchy: if $m \geq n$, then

$$\begin{aligned} d(f^m(x), f^n(x)) &= d(f^n(x), f^n(f^{m-n}(x))) \\ &\leq k^n d(x, f^{m-n}(x)) \\ &\leq \frac{k^n}{1 - k} d(x, f(x)) \end{aligned}$$

We know $k < 1$, so

$$\lim_{n \rightarrow \infty} k^n \left(\frac{d(x, f(x))}{1 - k} \right) = 0$$

So pick N s.t. the above is less than ε for all $n \geq N$. Then if $m \geq n \geq N$,

$$d(f^n(x), f^m(x)) \leq \frac{k^n}{1 - k} d(x, f(x)) < \varepsilon$$

So $(f^n(x))$ is Cauchy. So it converges to some x^* .

We claim that $f(x^*) = x^*$: since f is Lipschitz, f is continuous, and $f^n(x) \rightarrow x^*$, so $f(f^n(x)) \rightarrow f(x^*)$. But $f^{n+1}(x) \rightarrow x^*$. So $f(x^*) = x^*$.

We also claim that x^* is the only fixed point: Suppose $f(y) = y$. Then $d(f(x^*), f(y)) = d(x^*, y)$. But $d(f(x^*), f(y)) \leq kd(x^*, y)$, since f is a contraction where $k < 1$, this can only happen if $d(x^*, y) = 0$, i.e $x^* = y$. □

6 Solving Equations

Problem: Suppose $U \subset \mathbb{R}^n$ is open. $f : U \rightarrow \mathbb{R}^m$ is C^1 and $f(x_0) = y_0$. Can we solve $f(x) = y$ for y close to y_0 ?

If so, what does the set of x close to x_0 solution look like?

There are three cases:

a) $n < m$. For 'most' $y \in \mathbb{R}^m$, there is no solution. Idea: $\dim(U) \leq n < m$.

b) $m = n$. If y is sufficiently close to y_0 and $Df|_{x_0}$ is an isomorphism, then there is a unique solution near x_0 (inverse function theorem).

c) $m < n$. If $Df|_{x_0}$ is surjective and y is close to y_0 , set of solutions near x_0 looks like $B_\varepsilon(0) \subset \mathbb{R}^{n-m}$ (implicit function theorem).

We'll prove (b) and use it to prove (c).

6.1 Newton's method

$n = 1$: solve $f(x) = y^* = y$.

Approximate f by graph of it's tangent line at $(x_0, f(x_0))$.

$$g(x) = f(x_0) + f'(x_0)(x - x_0)$$

solve $g(x_1) = y^*$:

$$x_1 = x_0 + \frac{y^* - f(x_0)}{f'(x_0)}$$

Now repeat:

$$x_2 = x_1 + \frac{y^* - f(x_1)}{f'(x_1)}$$

and etc. Hope that $(x_n) \rightarrow x^*$ with $f(x^*) = y^*$.

General case: $f : U \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$: approximate $f(x)$ near $x = x_0$ by

$$g(x) = f(x_0) + Df|_{x_0}(x - x_0)$$

solve equation $g(x_1) = y$: $x_1 = x_0 + (Df|_{x_0})^{-1}(y - f(x_0))$.

Repeat: $x_2 = x_1 + (Df|_{x_1})^{-1}(y - f(x_1))$ etc.

Equivalently: define $n_y(x) = x + (Df|_x)^{-1}(y - f(x))$. Then

$$(x_k) = (n_y^k(x_0))$$

(the k^{th} iterate of n_y).

If x is a fixed point of n_y , then

$$\begin{aligned} x &= x + (Df|_x)^{-1}(y - f(x)) \\ \implies 0 &= (Df|_x)^{-1}(y - f(x)) \\ \implies 0 &= y - f(x) \\ \implies f(x) &= y \end{aligned}$$

so we have a solution.

So if we knew n_y was a contraction map, we would get a solution.

Problem: This only makes sense if $Df|_x$ is invertible. Analyzing $(Df|_x)^{-1}$ term is painful.

Modified Newton's method:

Suppose $f : U \rightarrow \mathbb{R}^n$ is C^1 , $f(x_0) = y_0$ and $Df|_{x_0} = A$ is invertible where $A : \mathbb{R}^n \rightarrow \mathbb{R}^n$ is a linear map.

Approximate $Df|_x$ by $Df|_{x_0} = A$, i.e. consider

$$N_y(x) = x + A^{-1}(y - f(x))$$

If $N_y(x) = x$, then $f(x) = y$ so we found a solution.

Is N_y a contraction map when x is close to x_0 ?

Compute

$$\begin{aligned} N_y(x) - N_y(x') &= x + A^{-1}(y - f(x)) - (x' + A^{-1}(y - f(x'))) \\ &= x - x' + A^{-1}(f(x') - f(x)) \\ &= A^{-1}(A(x) - f(x) - (A(x') - f(x'))) \\ &= A^{-1}(h(x) - h(x')) \end{aligned}$$

where $h(x) = A(x) - f(x)$.

Notice:

$$\begin{aligned} Dh|_x &= DA|_x - Df|_x \\ &= A|_x - Df|_x \\ &= Df|_{x_0} - Df|_x \end{aligned}$$

C^1 maps: $Df = U \rightarrow \mathcal{L}(\mathbb{R}^n, \mathbb{R}^n) = M_{n \times n}(\mathbb{R}) = \mathbb{R}^{n^2}$.

f is C^1 if Df is continuous.

Note: Since all norms on \mathbb{R}^{n^2} are equivalent, we can use whatever norm on $\mathcal{L}(\mathbb{R}^n, \mathbb{R}^n)$ we like.

For applications: use operator norm $\|\cdot\|_{op}$.

Lemma. If $\|Dh|_x\|_{op} < M$ for all $x \in B_r(x_0)$, then $\|h(x) - h(x')\|_2 \leq M\sqrt{n}\|x - x'\|$ for $x, x' \in B_r(x_0)$ (n is the dimension of space).

Proof. let h_i be the i^{th} component of h . Then

$$\|Dh_i|_x\| = \|Dh|_x(e_i)\| \leq M \cdot \|e_i\|_2 = M$$

So by Mean value inequality, $|h_i(x) - h_i(x')| \leq M \cdot \|x - x'\|_2$. So

$$\|h(x) - h(x')\| \leq \sqrt{n}M \cdot \|x - x'\|$$

□

Proposition. Given $\varepsilon > 0$, there exists $\delta > 0$ s.t. N_y is ε -lipschitz on $B_\delta(x_0)$.

Proof. f is C^1 , so choose $\delta > 0$ s.t.

$$\|Dh|_x\| = \|Df|_x - Df|_{x_0}\|_{op} \leq \frac{\varepsilon}{\|A^{-1}\|_{op} \cdot \sqrt{n}}$$

for $x \in B_\delta(x_0)$, so

$$\begin{aligned} \|N_y(x) - N_y(x')\| &= \|A^{-1}(h(x) - h(x'))\| \\ &\leq \|A^{-1}\|_{op} \|h(x) - h(x')\| \\ &\leq \|A^{-1}\|_{op} \cdot \sqrt{n}(\varepsilon/\sqrt{n} \cdot \|A^{-1}\|_{op}) \cdot (\|x - x'\|) \end{aligned}$$

□

6.2 The Inverse Function Theorem (See alternative notes)

Let $U \subset \mathbb{R}^n$ be open, $f(\mathbf{x}_0) = \mathbf{y}_0$, and $f : U \rightarrow \mathbb{R}^n$ is C^1 , $A = Df|_{\mathbf{x}_0}$ is invertible.

Let $a = \|A^{-1}\|_{op}$, so that

$$\|A^{-1}(v)\| \leq a\|v\|$$

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

Lemma. $\exists n > 0$ s.t. $Df|_{\mathbf{x}}$ is invertible for all $\mathbf{x} \in B_n(\mathbf{x}_0)$.

Proof. $f : U \rightarrow \mathbb{R}^n$ is C^1 , so the map

$$\begin{array}{ccc} \alpha : U & \rightarrow \mathcal{L}(\mathbb{R}^n, \mathbb{R}^n) & \rightarrow \mathbb{R} \\ \mathbf{x} & \rightarrow Df|_{\mathbf{x}} & \rightarrow \det(Df|_{\mathbf{x}}) \end{array}$$

is continuous.

$\mathbb{R} - \{0\}$ is open in \mathbb{R} , so $\alpha^{-1}(\mathbb{R} - \{0\})$ is open in U .

$\alpha^{-1}(\mathbb{R} - \{0\}) = \{\mathbf{x} \in U \mid Df|_{\mathbf{x}} \text{ is invertible}\}$. $\mathbf{x}_0 \in \alpha^{-1}(\mathbb{R} - \{0\})$, so $\exists n > 0$ s.t. $B_n(\mathbf{x}_0) \subset \alpha^{-1}(\mathbb{R} - \{0\})$. □

Consider $N_{\mathbf{y}}(\mathbf{x}) = \mathbf{x} + A^{-1}(\mathbf{y} - f(\mathbf{x}))$ (Modified Newton's Method).

Fix r_0 s.t. $0 < r_0 < n$ and $N_{\mathbf{y}}$ is $\frac{1}{2}$ -Lipschitz on $\bar{B}_{r_0}(\mathbf{x}_0)$. By the lemma, $Df|_{\mathbf{x}}$ is invertible for $\mathbf{x} \in \bar{B}_{r_0}(\mathbf{x}_0)$.

We want $N_{\mathbf{y}}$ to be a contraction map.

Problem: $N_{\mathbf{y}}(B_{r_0}(\mathbf{x}_0))$ need not be in the region where $N_{\mathbf{y}}$ is contracting.

Solution: require \mathbf{y} to be close to \mathbf{y}_0 .

Proposition. (2) Let $r(\mathbf{y}) = 2a\|\mathbf{y} - \mathbf{y}_0\|$. If $r(\mathbf{y}) \leq r_0$, then $N_{\mathbf{y}} : \bar{B}_{r(\mathbf{y})}(\mathbf{x}_0) \rightarrow \bar{B}_{r(\mathbf{y})}(\mathbf{x}_0)$.

Proof. Suppose $\mathbf{x} \in B_{r(\mathbf{y})}(\mathbf{x}_0)$. Then

$$\begin{aligned} \|N_{\mathbf{y}}(\mathbf{x}_0)\| &\leq \|N_{\mathbf{y}}(\mathbf{x}) - N_{\mathbf{y}}(\mathbf{x}_0)\| + \|N_{\mathbf{y}}(\mathbf{x}_0) - \mathbf{x}_0\| \\ &\leq \frac{1}{2}\|\mathbf{x} - \mathbf{x}_0\| + \|A^{-1}(\mathbf{y} - \mathbf{y}_0)\| \\ &\leq \frac{1}{2}r(\mathbf{y}) + a\|\mathbf{y} - \mathbf{y}_0\| \\ &= \frac{1}{2}r(\mathbf{y}) + \frac{1}{2}r(\mathbf{y}) \\ &= r(\mathbf{y}) \end{aligned}$$

So $N_{\mathbf{y}}(\mathbf{x}_0) \in B_{r(\mathbf{y})}(\mathbf{x}_0)$. □

Proposition. (3) Suppose $r \leq r_0$. If $\mathbf{y} \in B_{\frac{r}{2a}}(\mathbf{y}_0)$, then there is a unique $\mathbf{x} \in B_r(\mathbf{x}_0)$ s.t. $f(\mathbf{x}) = \mathbf{y}$.

Proof. $f(\mathbf{x}) = \mathbf{y} \iff N_{\mathbf{y}}(\mathbf{x}) = \mathbf{x}$.

$N_{\mathbf{y}}$: If $\mathbf{y} \in \bar{B}_{r/2a}(\mathbf{y}_0)$, then $r(\mathbf{y}) = r$, so by the previous proposition,

$$N_{\mathbf{y}} : \bar{B}_r(\mathbf{x}_0) \rightarrow \bar{B}_r(\mathbf{x}_0)$$

$r \leq r_0$, so $N_{\mathbf{y}}$ is $\frac{1}{2}$ -Lipschitz on $\bar{B}_r(\mathbf{x}_0)$, i.e. $N_{\mathbf{y}} : \bar{B}_r(\mathbf{x}_0) \rightarrow \bar{B}_r(\mathbf{x}_0)$ is a contraction.

$\bar{B}_r(\mathbf{x}_0)$ is a closed subset of \mathbb{R}^n , which is complete, so $\bar{B}_r(\mathbf{x}_0)$ is complete. Thus there is a unique $\mathbf{x} \in \bar{B}_r(\mathbf{x}_0)$ such that $N_{\mathbf{y}}(\mathbf{x}) = \mathbf{x}$, i.e. there exists a unique $\mathbf{x} \in \bar{B}_r(\mathbf{x}_0)$ s.t. $f(\mathbf{x}) = \mathbf{y}$. □

Note: $r(\mathbf{y}) = 2a\|\mathbf{y} - \mathbf{y}_0\|$, so $r(\mathbf{y}) \leq r \implies \mathbf{y} \in \bar{B}_{r/2a}(\mathbf{y}_0)$.

Remark. If $\mathbf{y} \in \bar{B}_{r/2a}(\mathbf{y}_0)$ for $r < r_0$, then it's in $\bar{B}_{r_0/2a}(\mathbf{y}_0)$, so the proposition implies that there is a unique $\mathbf{x} \in \bar{B}_{r_0/2a}(\mathbf{y}_0)$ with $f(\mathbf{x}) = \mathbf{y}$ and $\mathbf{x} \in B_r(\mathbf{x}_0)$.

Proposition. (4) There are open sets $V \subset U$, $\mathbf{x}_0 \in V$, $W \subset \mathbb{R}^n$, $\mathbf{y}_0 \in W$ s.t. $f|_V : V \rightarrow W$ bijectively.

Proof. Take $W = B_{r_0/4a}(\mathbf{y}_0)$. f is continuous, so $f^{-1}(W)$ is open. Take

$$]V = f^{-1}(W) \cap B_{r_0}(\mathbf{x}_0)$$

which is open.

Given $\mathbf{y} \in W$, there exists $x \in \bar{B}_{r_0/2}(\mathbf{x}_0)$ with $f(\mathbf{x}) = y$ by the previous proposition. Moreover, this \mathbf{x} is the unique \mathbf{x} in that open ball. So $\mathbf{x} \in V = f^{-1}(W) \cap B_r(\mathbf{x}_0)$ and is the unique such element. \square

Proposition. (5) Let $g : W \rightarrow V$ be the inverse of f . Then g is continuous at \mathbf{y}_0 .

Proof. If $\mathbf{y} \in \bar{B}_{r/2a}(\mathbf{y}_0)$, then $g(\mathbf{y}) \in \bar{B}_r(\mathbf{x}_0)$ by proposition 3, i.e. if $\|\mathbf{y} - \mathbf{y}_0\| \leq \delta$, then $\|g(\mathbf{y}) - g(\mathbf{y}_0)\| \leq 2a\delta$. So g is continuous at \mathbf{y}_0 . \square

Proposition. (6) g is differentiable at \mathbf{y}_0 and $Dg|_{\mathbf{y}_0} = A^{-1}$.

Proof. Note that $g(\mathbf{y})$ satisfies $N_{\mathbf{y}}(g(\mathbf{y})) = g(\mathbf{y})$, so if $\mathbf{y} \in \bar{B}_{r/2a}(\mathbf{y}_0)$, $g(\mathbf{y}) \in N_{\mathbf{y}}(\bar{B}_{r(\mathbf{y})}(\mathbf{x}_0))$.

Now $N_{\mathbf{y}} : \bar{B}_{r(\mathbf{y})}(\mathbf{x}_0) \rightarrow \bar{B}_{r(\mathbf{y})}(\mathbf{x}_0)$ is $\varepsilon(r(\mathbf{y}))$ – Lipschitz, by proposition 1 where $\varepsilon(r(\mathbf{y})) \rightarrow 0$ as $r(\mathbf{y}) \rightarrow 0$.

So $N_{\mathbf{y}}(\bar{B}_{r(\mathbf{y})}(\mathbf{x}_0)) \subset B_{\varepsilon(r(\mathbf{y})) \cdot r(\mathbf{y})}(N_{\mathbf{y}}(\mathbf{x}_0))$, i.e. $g(\mathbf{y}) = N_{\mathbf{y}}(\mathbf{x}_0) + E(\mathbf{y})$, where

$$\begin{aligned} \|E(\mathbf{y})\| &\leq \varepsilon(r(\mathbf{y})) \cdot r(\mathbf{y}) \\ &= \varepsilon(r(\mathbf{y}))2a\|\mathbf{y} - \mathbf{y}_0\| \\ &= \mathbf{x}_0 + A^{-1}(\mathbf{y} - \mathbf{y}_0) + E(\mathbf{y}) \end{aligned}$$

where

$$\frac{\|E(\mathbf{y})\|}{\|\mathbf{y} - \mathbf{y}_0\|} \leq 2a\varepsilon(r(\mathbf{y}))$$

and $\varepsilon(r(\mathbf{y})) \rightarrow 0$ as $\|\mathbf{y} - \mathbf{y}_0\| \rightarrow 0$. So the above equation says that g is differentiable at \mathbf{y}_0 , and $Dg|_{\mathbf{y}_0} = A^{-1}$. \square

Definition. Suppose $V, W \subset \mathbb{R}^n$ are open. $f : V \rightarrow W$ is a *diffeomorphism* if

- f is bijective;
- f and f^{-1} are both C^1 .

Theorem. (Inverse function theorem) Suppose $U \subset \mathbb{R}^n$ is open, $f : U \rightarrow \mathbb{R}^n$ is C^1 with $f(\mathbf{x}_0) = \mathbf{y}_0$ and $Df|_{\mathbf{x}_0}$ is invertible. Then there are open subsets $V \subset U$ and $\mathbf{x}_0 \in V$, $W \subset \mathbb{R}^n$ and $\mathbf{y}_0 \in W$ s.t. $f|_V : V \rightarrow W$ is a diffeomorphism.

Proof. Let V and W be as in Proposition 4. Then $f : V \rightarrow W$ bijectively. Let $g = f^{-1} : W \rightarrow V$. Must show g is C^1 .

We know $V \subset B_{r_0}(\mathbf{x}_0)$ where $Df|_{\mathbf{x}}$ is invertible for all $\mathbf{x} \in B_r(\mathbf{x}_0)$ (hypothesis of this subsection).

Apply proposition 6 with \mathbf{x} in place of \mathbf{x}_0 , we see that g is differentiable at \mathbf{x} , and $Dg|_{\mathbf{x}} = (Df|_{\mathbf{x}})^{-1}$.

g is differentiable implies that g is continuous. To see g is C^1 , note that $Dg : W \rightarrow \mathcal{L}(\mathbb{R}^n, \mathbb{R}^n)$ is a composition

$$\begin{array}{ccccccc} W & \rightarrow & V & \rightarrow & \mathcal{L}(\mathbb{R}^n, \mathbb{R}^n) & \rightarrow & \mathcal{L}(\mathbb{R}^n, \mathbb{R}^n) \\ \mathbf{y} & \rightarrow & g(\mathbf{y}) & & & & \\ & & & x \rightarrow & Df|_{\mathbf{x}} & & \\ & & & & A \rightarrow & & A^{-1} \end{array}$$

□

Reformulation: Local change of coordinates:

Suppose $U_i = f_i(x_1, \dots, x_n)$ for $1 \leq i \leq n$.

Consider $J = \left(\frac{\partial u_i}{\partial x_j} \right) = (D_j f_i) =$ matrix representing Df .

If $\det(J|_{\mathbf{x}_0}) \neq 0$ (i.e. $Df|_{\mathbf{x}_0}$) then we can use (U_1, \dots, U_n) as a local system of coordinates near \mathbf{x}_0 .

i.e. we can solve for x_j 's in terms of u_i 's:

$$x_j = g_j(u_1, \dots, u_n).$$

Example. Polar coordinates: $x = r \cos \theta$, $y = r \sin \theta$,

$$J = \begin{pmatrix} \cos \theta & -r \sin \theta \\ \sin \theta & r \cos \theta \end{pmatrix}$$

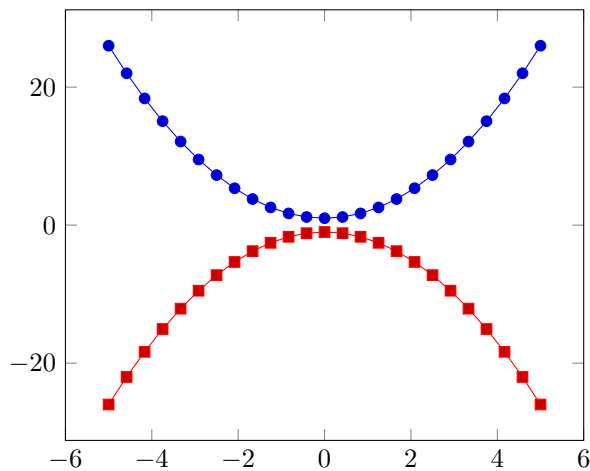
$\det J = r$ i.e. there's a good change of coordinates between polar and rectangular coordinates except when $r = 0$.

6.3 The implicit function theorem

Let $F : \mathbb{R}^n \rightarrow \mathbb{R}^m$ be C^1 ($n \geq m$), $F(\mathbf{x}_0) = \mathbf{y}_0$.

Problem: Describe $F^{-1}(\mathbf{y}_0)$ near \mathbf{x}_0 .

Example. $F : \mathbb{R}^2 \rightarrow \mathbb{R}$, $F(x, y) = x^2 - y^2$.



Notation. $B_\varepsilon^k = B_\varepsilon(\mathbf{0}) \subset \mathbb{R}^k = k$ -dimensional open ball.

Theorem. Suppose $F: \mathbb{R}^n \rightarrow \mathbb{R}^m$ is C^1 , $f(\mathbf{x}_0) = \mathbf{y}_0$, and $DF|_{\mathbf{x}_0}$ is surjective. Then there's an open set $V \subset \mathbb{R}^n$, $\mathbf{x}_0 \in V$, and a C^1 map $G: B_\varepsilon^{n-m} \rightarrow \mathbb{R}^n$ such that:

1) $F^{-1}(\mathbf{y}_0) \cap V = \text{im } G$;

2) G is injective;

3) $DG|_{\mathbf{z}}$ is injective for all $\mathbf{z} \in B_\varepsilon^{n-m}$.

i.e. if $n - m = 1$, $B'_\varepsilon = (-\varepsilon, \varepsilon)$, $F^{-1}(\mathbf{y}_0) \cap V$ is a parametrized curve;

if $n - m = 2$ then this is a parametrized surface.

For general $n - m$, we call this is a parametrized $(n - m)$ -manifold.

Example. $F(x, y) = x^2 - y^2$, $DF|_{(x, y)} = [2x, -2y]$ is surjective $\iff (x, y) \neq (0, 0)$.

Definition. $F^{-1}(\mathbf{y}_0)$ is *smooth* at \mathbf{x}_0 if $DF|_{\mathbf{x}_0}$ is surjective, *singular* at \mathbf{x}_0 otherwise.

$F^{-1}(\mathbf{y}_0)$ is smooth if it is smooth at all $\mathbf{x} \in F^{-1}(\mathbf{y}_0)$.

Proof of theorem:

Proof. $DF|_{\mathbf{x}_0}: \mathbb{R}^n \rightarrow \mathbb{R}^m$ is surjective. So $K := \ker DF|_{\mathbf{x}_0}$ has dimension $(n - m)$. Choose any $\pi \in \mathcal{L}(\mathbb{R}^n, \mathbb{R}^m)$ with $\pi(K) = \mathbb{R}^{n-m}$. Define $f: \mathbb{R}^n \rightarrow \mathbb{R}^m \times \mathbb{R}^{n-m} = \mathbb{R}^n$ by $f(\mathbf{x}) = (F(\mathbf{x}), \pi(\mathbf{x}))$. So $Df: \mathbb{R}^n \rightarrow \mathbb{R}^m \oplus \mathbb{R}^{n-m}$.

$Df_{\mathbf{x}_0}(\mathbf{v}) = (DF|_{\mathbf{x}_0}(\mathbf{v}), \pi(\mathbf{v}))$ since $D\pi = \pi$.

Claim: $Df|_{\mathbf{x}_0}$ is an isomorphism: If $Df|_{\mathbf{x}_0}(\mathbf{v}) = \mathbf{0}$, then $DF|_{\mathbf{x}_0}(\mathbf{v}) = \mathbf{0} \implies \mathbf{v} \in K$. But $\pi: K \rightarrow \mathbb{R}^{n-m}$ is an isomorphism, so $\pi(\mathbf{v}) = \mathbf{0} \implies \mathbf{v} = \mathbf{0}$. So $\ker Df|_{\mathbf{x}_0} = \{\mathbf{0}\} \implies Df|_{\mathbf{x}_0}$ is an isomorphism.

By the inverse function theorem, there exists $V \subset \mathbb{R}^n$, $\mathbf{x}_0 \in V$, $W \subset \mathbb{R}^m \times \mathbb{R}^{n-m}$, $(\mathbf{y}_0, \pi(\mathbf{x}_0)) \in W$, s.t. $f: V \rightarrow W$ is a diffeomorphism. Let $g = f^{-1}: W \rightarrow V$.

Then $F^{-1}(\mathbf{y}_0) \cap V = f^{-1}(\mathbf{y}_0 \times \mathbb{R}^{n-m}) \cap V$, so $g(\mathbf{y}_0 \times \mathbb{R}^{n-m}) \cap W = F^{-1}(\mathbf{y}_0) \cap V$.

Define $G(\mathbf{z}) = g(\mathbf{y}_0, \mathbf{z}_0)$, g is injective implies that G is injective, and D_g injective $\implies DG$ injective. \square