

Advanced Calculus

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February 26, 2026

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1 Metric Spaces

Definition 1.1. Let X be a set, a *metric* d is a map

$$d: X \times X \rightarrow \mathbb{R}$$

such that:

- (i) $d(x, y) \geq 0$ with equality iff $x = y$.
- (ii) $d(x, y) = d(y, x)$
- (iii) $d(x, z) \leq d(x, y) + d(y, z)$

Example 1.2. Given a vector space V and norm $\|\cdot\|$ we can define the metric

$$d(x, y) = \|x - y\|$$

Question. Given a vector space V and a metric d on V , does d "come from" a norm? i.e does there exist some norm $\|\cdot\|$ s.t

$$d(x, y) = \|x - y\|$$

The answer in general is no (consider the discrete metric, or even any bounded metric: this contradicts homogeneity).

Notation. Take (X, d) metric space $\delta > 0$ and $x_0 \in X$, the δ -neighborhood of x_0

$$N_\delta(x_0) := \{x \in X : d(x, x_0) < \delta\}$$

Definition 1.3. (X, d) metric space and $E \subset X$

(i) $x_0 \in X$ is an *interior point* of E iff $\exists \delta > 0$ s.t

$$N_\delta(x_0) \subset E$$

(ii) $x_0 \in X$ is a *limit point* iff $\forall \delta > 0$

$$N_\delta^*(x_0) \cap E \neq \emptyset$$

(We denote by E' the set of limit points of E)

(iii) E is *open* iff every $x_0 \in E$ is an interior point of E .

(iv) E is *closed* iff every $E' \subset E$.

(v) The *closure* of E denoted \overline{E} is

$$\overline{E} := E \cup E'$$

Review. Fix (X, d) a metric space.

1. $N_\delta(x_0)$ is open $\forall \delta > 0, \forall x_0 \in X$.
2. E is open iff E^c is closed.
3. \overline{E} is closed, in fact \overline{E} is the smallest closed set containing E .
4. The union of any collection of open sets is open.
5. A finite intersection of a collection of open sets is open.
6. Intersection of any collection of closed sets is closed.
7. Finite union of closed is closed.
8. If $x \in E'$ then $N_\delta(x) \cap E'$ is an infinite set.

Definition 1.4. Let x_n be a sequence in X , we say that x_n converges to $x_0 \in X$ iff $\forall \varepsilon > 0 \exists N \in \mathbb{N}$ s.t

$$d(x_n, x_0) < \varepsilon \quad \forall n \geq N$$

1. The limit of a sequence (if exists) is unique.
2. x_n converges to x_0 in (X, d) iff $d(x_n, x_0)$ converges 0 in $(\mathbb{R}, ||\cdot||)$.

Example 1.5. $x_n = (1/n + 1, e^{-n})$ in \mathbb{R}^2 . Find the limit of x_n in $(\mathbb{R}^2, \|\cdot\|_\infty)$, $(\mathbb{R}^2, \|\cdot\|_1)$, $(\mathbb{R}^2, \|\cdot\|_2)$.

•

$$\|(1/n + 1, e^{-n}, (1, 0))\|_\infty = \max(1/n, e^{-n}) \leq 1/n + e^{-n} \rightarrow 0$$

• $x_0 = (1, 0)$

$$\|(1/n + 1, e^{-n}, -(1, 0))\|_1 = \|(1/n, e^{-n})\|_1 = |1/n| + |e^{-n}| \rightarrow 0$$

•

$$\|(1/n + 1, e^{-n}) - (1, 0)\|_2 = \sqrt{1/n^2 + e^{-2n}} \rightarrow 0$$

Remark 1.6. $x_n \rightarrow x_0$ in $(\mathbb{R}^k, \|\cdot\|_\infty)$

$$(x_{1,n}, x_{2,n}, \dots, x_{k,n}) \rightarrow (x_1, 0, \dots, x_k, 0)$$

Iff

$$x_{1,n} \rightarrow x_{1,0}$$

$$x_{2,n} \rightarrow x_{2,0}$$

$$\vdots$$

$$x_{k,n} \rightarrow x_{k,0}$$

in $(\mathbb{R}, \|\cdot\|)$.

Proof. $|x_{i,n} - x_{i,0}| \leq \|x_n - x_0\|_\infty \leq |x_{1,n} - x_{1,0}| + \dots + |x_{k,n} - x_{k,0}|$ Therefore $x_n \rightarrow x_0$ in $(\mathbb{R}^k, \|\cdot\|_\infty)$
 $\iff \|x_n - x_0\|_\infty \rightarrow 0$ and $\iff |x_{i,n} - x_{i,0}| \rightarrow 0 \forall i$ and hence $\iff x_{i,n} \rightarrow x_{i,0} \forall i$. \square

Remark 1.7. Take $x \in \mathbb{R}^n$, then

$$\|x\|_\infty = \max_i(x_i) \leq |x_1| + \dots + |x_n| = \|x\|_1 \leq n\|x\|_\infty$$

Therefore

$$\|x\|_\infty \leq \|x\|_1 \leq n \cdot \|x\|_\infty$$

Similarly, show something for $\|x\|_2$ and $\|x\|_\infty$.

Definition 1.8. Given a vector space V and norms $\|\cdot\|_a$ and $\|\cdot\|_b$ we say that $\|\cdot\|_a$ and $\|\cdot\|_b$ are equivalent iff $\exists c_1, c_2 > 0 \in \mathbb{R}$ s.t $\forall x \in V$

$$c_1\|x\|_b \leq \|\cdot\|_a \leq c_2\|x\|_b$$

Lemma 1.9. (X, d) metric space and $E \subseteq X$, then E is closed iff for sequence $x_n \in E$ that converges $\lim_{n \rightarrow \infty} x_n \in E$.

Proof. \implies Assume that there exists a sequence x_n in E that converges to x_0 s.t $x_0 \notin E$, take $\varepsilon > 0$, then since $x_n \rightarrow x_0$ then $\exists N \in \mathbb{N}$ s.t

$$x_n \in N_\varepsilon(x_0) \quad \forall n \geq N$$

But $x_n \in E$,

$$N_\varepsilon^*(x_0) \cap E \neq \emptyset$$

Therefore

$$x_0 \in E'$$

Contradiction to the fact that E is closed.

- Take $x_0 \in E'$, let $\varepsilon_n = 1/n$ then

$$\exists x_n \in N_{\varepsilon_n}^*(x_0) \cap E$$

Consider $\{x_n\}_{n \in \mathbb{N}}$ is a sequence in E s.t

$$d(x_n, x_0) < 1/n \implies x_n \rightarrow x_0$$

and hence $x_0 \in E$.

□

Theorem 1.10. Given a vector space V and norms $\|\cdot\|_a$ and $\|\cdot\|_b$ then the following are all equivalent:

1. $\|\cdot\|_a$ and $\|\cdot\|_b$ are equivalent ($c_1\|x\|_b \leq \|x\|_a \leq c_2\|x\|_b$).
2. $x_n \rightarrow x_0$ in $\|\cdot\|_a \iff x_n \rightarrow x_0$ in $\|\cdot\|_b$.
3. E is closed in $(V, \|\cdot\|_a) \iff E$ is closed in $(V, \|\cdot\|_b)$.
4. U is open in $(V, \|\cdot\|_a) \iff U$ is open in $(V, \|\cdot\|_b)$.

Proof. $\implies 2$

$$c_1\|x_n - x_0\|_b \leq \|x_n - x_0\|_a \leq c_2\|x_n - x_0\|_b \leq \frac{c_2}{c_1}\|x_n - x_0\|_a$$

Apply Squeeze.

2 \implies 3 By the Lemma.

3 \implies 4 Trivial.

4 \implies 1 $B_{\|\cdot\|_a}(0, 1)$ open in $(V, \|\cdot\|_a)$ therefore it is open in $(V, \|\cdot\|_b)$. Hence 0 is an interior point in $(V, \|\cdot\|_b)$. Hence $\exists \delta > 0$

$$B_{\|\cdot\|_b}(0, \delta) \subseteq B_{\|\cdot\|_a}(0, 1)$$

Let $x \in V, x \neq 0$.

$$\frac{\delta}{2} \cdot \frac{x}{\|x\|_b} \in B_{\|\cdot\|_b}(0, \delta) \subseteq B_{\|\cdot\|_a}(0, 1)$$

Therefore

$$\left\| \frac{\delta}{2} \frac{x}{\|x\|_b} \right\|_a < 1$$

We get

$$\frac{\delta}{2\|x\|_b} \|x_a\| < 1 \implies \|x\|_a \leq \frac{2}{\delta} \|x\|_b \quad \forall x \in V$$

□

1.1 Compact Sets

Definition 1.11. (X, d) metric space, $K \subseteq X$. We say that K is *compact* iff for every open cover

$$\mathcal{G} = \{G_\alpha\}_{\alpha \in I}$$

of K has a finite subcover.

Example 1.12. $(0, 1]$ is not compact in $(\mathbb{R}, |\cdot|)$, take

$$\mathcal{G} = \{(1/n, 10)\}_{n \in \mathbb{N}}$$

which has no finite subcover.

Theorem 1.13 (Heine-Borel). *In \mathbb{R} $[a, b]$ is compact.*

Example 1.14. If X is finite, then X is compact.

Proof. Let $X = \{x_1, \dots, x_n\}$ and let

$$\mathcal{G} = \{G_\alpha\}_{\alpha \in I}$$

be an open cover X . Then for every x_i , fix some G_{α_i} s.t

$$x = \{x_1, \dots, x_n\} \subseteq \bigcup_{i=1}^n G_{\alpha_i}$$

□

Proposition 1.15. (X, d) metric space and $K \subseteq X$ then every closed subset E of K is compact.

Proof. Let $\mathcal{G} = \{G_\alpha\}_{\alpha \in I}$ be an open cover of E , then

$$E \subseteq \bigcup_{\alpha \in I} G_\alpha$$

Therefore

$$K \subseteq \bigcup G_\alpha \cup E^c, \quad E^c \text{ open}$$

Therefore

$$\mathcal{G}' = \mathcal{G} \cup \{E^c\} \quad \text{is an open cover of } K$$

Therefore it has a finite subcover and

$$K \subseteq G_{\alpha_1} \cup \dots \cup G_{\alpha_k} \cup E^c$$

Hence

$$E \subseteq G_{\alpha_1} \cup \cdots \cup G_{\alpha_k}$$

Which is a finite subcover of \mathcal{G} . □

Definition 1.16. (X, d) metric space and x_n sequence in X and

$$\varphi: \mathbb{N} \rightarrow \mathbb{N} \quad \text{strictly increasing}$$

We say that $b_n = a_{\varphi(n)}$ is a *subsequence* of a_n .

Proposition 1.17. x_n converges to $x_0 \iff$ every subsequence x_{n_k} converges to x_0 .

Definition 1.18. (X, d) metric space and $K \subseteq X$. We say that K is sequentially compact iff every sequence $x_n \in K$ has a convergent subsequence in K .

Proposition 1.19. (X, d) metric space, K sequentially compact iff every infinite $E \subseteq K$ has a limit point in K .

Proof. \implies Let K be sequentially compact, take $E \subseteq K$ an infinite set. We can extract from E a sequence of distinct elements $x_n \in E \subseteq K$. Therefore x_n has a convergence subsequence x_{n_k} s.t the limit of $x_{n_k} = x_0 \in K$. Moreover, x_{n_k} are distinct in E therefore $x_0 \in E' \cap K$.
 \impliedby Let x_n be a sequence in K , take

$$E = \{x_n\}_{n \in \mathbb{N}}$$

We have two cases:

- **E is finite.** Then $\exists x_0$ and $\exists n_1 < n_2 < \cdots$ (infinitely many) s.t

$$x_{n_k} = x_0, \quad \forall k \in \mathbb{N}$$

Therefore

$$\lim_{k \rightarrow \infty} x_{n_k} = x_0 \in E$$

We found a converging subsequence.

- **E is infinite.** Then by the assumption, we have that E has a limit point x_0 in K .

$$\varepsilon = 1 \quad \exists n_1 \text{ s.t } x_{n_1} \in N_1^*(x_0) \cap E.$$

$$\varepsilon = 1/2 \quad \exists n_2 > n_1 \text{ s.t } x_{n_2} \in N_{1/2}^*(x_0) \cap E \text{ (since } N_{1/2}^*(x_0) \cap E \text{ is infinite).}$$

we can construct $n_k > n_{k-1}$ s.t $x_{n_k} \in N_{1/k}^*(x_0) \cap E$, therefore

$$\lim_{n \rightarrow \infty} x_{n_k} = x_0 \in K$$

□

Example 1.20. In $(\mathbb{Q}, |\cdot|)$, $K = \mathbb{Q} \cap [0, 1]$ is closed and bounded. P_n increasing sequence in $\mathbb{Q} \cap [0, 1]$ s.t

$$P_n \rightarrow 1/\sqrt{2} \in \mathbb{R}$$

Consider

$$\mathcal{G} = \left\{ (-1, P_1) \cap \mathbb{Q}, (-1, P_2) \cap \mathbb{Q}, \dots, (-1, P_n) \cap \mathbb{Q}, \dots, (1/\sqrt{2}, 10) \cap \mathbb{Q} \right\}$$

Proposition 1.21. $K \subseteq X$ is compact K is sequentially compact.

Proof. • Let $E \subseteq K$ infinite, suppose that $E' \cap K = \emptyset$. Take $x \in K \implies x \notin E'$, therefore $\exists \delta_x > 0$ s.t

$$N_{\delta_x}^*(x) \cap E = \emptyset$$

Let $\mathcal{G} = \{N_{\delta_x}(x)\}_{x \in K}$, is an open cover of K compact, hence it has a finite subcover

$$K \subseteq N_{\delta_{x_1}}(x_1) \cup \dots \cup N_{\delta_{x_k}}(x_k)$$

But $E \subset K \implies E = (E \cap K)$ hence

$$\begin{aligned} E &\subseteq (N_{\delta_{x_1}}(x_1) \cap E) \cup \dots \cup (N_{\delta_{x_k}}(x_k) \cap E) \\ &\subseteq \{x_1, \dots, x_k\} \end{aligned}$$

therefore E is finite, contradiction. □

1.2 Complete sets

Definition 1.22. Given a sequence x_n , we say that x_n is Cauchy iff for every $\varepsilon > 0$ there exists $N \in \mathbb{N}$ s.t

$$d(x_n, x_m) < \varepsilon \quad \text{for } n, m \geq N$$

Proposition 1.23. If x_n converges, then x_n is Cauchy.

Proof. Triangle inequality. □

Remark 1.24. The converse is false. In $(\mathbb{Q}, |\cdot|)$, in $(\mathbb{R}, |\cdot|)$ there exists some sequence $x_n \in \mathbb{Q}$ s.t

$$x_n \rightarrow \sqrt{2} \quad \text{as } n \rightarrow \infty$$

But $x_n \rightarrow \sqrt{2}$ implies that x_n is Cauchy in $(\mathbb{R}, |\cdot|)$, therefore x_n is Cauchy in \mathbb{Q} . Suppose x_n converges to P in $(\mathbb{Q}, |\cdot|)$, therefore $x_n \rightarrow p \in (\mathbb{R}, |\cdot|)$ by uniqueness of limit, $p = \sqrt{2}$, contradiction.

Definition 1.25. (X, d) metric space and $K \subseteq X$, we say that K is complete iff every Cauchy Sequence in K converges in K .

Example 1.26. $(\mathbb{R}, |\cdot|)$ is complete (Math 210)

Proposition 1.27. If K is complete then K is closed.

Proof. Let x_n be a sequence in K s.t $x_n \rightarrow x_0 \in X$. Since x_n converges, then x_n is Cauchy and K is complete then x_n converges in K , therefore $x_0 \in K$. □

Proposition 1.28. K complete and $E \subseteq K$ is closed, then E is complete.

Proof. Let x_n be a Cauchy sequence in E , then x_n is Cauchy in K which is complete, therefore $x_n \rightarrow x_0 \in K$. But since E is closed, we get that $x_0 \in E$. □

Proposition 1.29. (X, d) metric space and x_n Cauchy sequence in $K \subseteq X$. If x_n has a convergent subsequence x_{n_k} then x_n converges to the subsequential limit.

Proof. Triangle inequality. □

Proposition 1.30. (X, d) metric space and $K \subseteq X$ is sequentially compact then K is complete (and therefore closed).

Proof. Let $\{x_n\}$ be a Cauchy sequence in K , K is sequentially compact therefore x_n has a convergent subsequence x_{n_k} that converges in K . Therefore by the previous proposition we get that x_n converges in K . □

Theorem 1.31. Let V be a finite dimensional inner product space over \mathbb{R} , then V is complete w.r.t norm induced by the inner product. Therefore $K \subseteq V$ is complete $\iff K$ is closed.

(For the converse $K \subseteq V$ which is closed, then K is complete).

Proof. Let $S = \{v_1, \dots, v_l\}$ be an orthonormal basis of V , take x_n be a cauchy sequence in V . Therefore

$$x_n = r_{1,n}v_1 + \dots + r_{l,n}v_l$$

Take $r_{1,n} = \langle x_n, v_1 \rangle$ therefore

$$\begin{aligned} |r_{1,m} - r_{1,n}| &= |\langle x_m, v_1 \rangle - \langle x_n, v_1 \rangle| \\ &= |\langle x_m - x_n, v_1 \rangle| \\ &\leq \|x_m - x_n\| \|v_1\| \quad \text{By C.S} \end{aligned}$$

Let $\varepsilon > 0$ then $\exists N \in \mathbb{N}$ s.t

$$\|x_m - x_n\| < \varepsilon \implies r_{1,n} \text{ is Cauchy in } \mathbb{R}$$

Therefore $r_{1,n}$ converges to $r_1 \in \mathbb{R}$. Do the same for every coordinate, and define

$$x = r_1v_1 + \dots + r_nv_n$$

Therefore

$$\begin{aligned} \|x_n - x\| &= \|(r_{1,n} - r_1)v_1 + \dots + (r_{l,n} - r_l)v_l\| \\ &\leq \|r_{1,n} - r_1\| \|v_1\| + \dots + \|r_{l,n} - r_l\| \|v_l\| \end{aligned}$$

Therefore

$$x_n \rightarrow x$$

□

1.3 Total Boundedness

Definition 1.32. (X, d) metric space and $K \subseteq X$ we say that K is totally bounded iff for every $\varepsilon > 0 \exists x_1, \dots, x_t \in K$ s.t

$$K \subseteq N_\varepsilon(x_1) \cup \dots \cup N_\varepsilon(x_t)$$

Definition 1.33. $K \subseteq X$ is bounded iff $\exists R > 0$ s.t

$$K \subseteq N_R(x_0) \quad \text{for some } x_0 \in K$$

We say that a sequence is bounded iff

$$E = \{x_n\}_{n \in \mathbb{N}} \quad \text{is bounded}$$

Proposition 1.34. *Cauchy sequences are bounded.*

Proposition 1.35. *$K \subseteq X$ is totally bounded it is bounded.*

Note that the converse is false.

Example 1.36. $(\mathbb{R}, d_{disc}), \mathbb{R}$ is bounded since

$$\mathbb{R} \subseteq N_2(0)$$

but it isn't totally bounded, take $\varepsilon = 1/2$ then if it is totally bounded, then $\exists x_1 \dots x_t \in \mathbb{R}$ s.t

$$\mathbb{R} \subseteq N_{1/2}(x_1) \cup \dots \cup N_{1/2}(x_t) = \{x_1, \dots, x_t\}$$

Proof. Take $\varepsilon = 1$ then $K \subseteq N_1(x_1) \cup N_1(x_t)$ for $x_1, \dots, x_t \in K$. Take x_1 and let $d = \max_{k \in \{1, \dots, t\}}(x_1, x_t)$, therefore for any $x \in K$ we have that

$$d(x, x_1) \leq d(x, x_i) + d(x_i, x_1)$$

for some x_i s.t $x \in N_1(x_i)$ (Exists by the above), therefore

$$d(x, x_1) \leq 1 + d \implies x \in N_{1+d}(x_1)$$

□

Proposition 1.37. *K is sequentially compact then K is totally bounded.*

Proof. Suppose not, then $\exists \varepsilon_0$ s.t K cannot be covered by finitely many ε_0 -neighborhoods. Fix some x_1 then $\exists x_2 \in K \setminus N_{\varepsilon_0}(x_1)$, Similarly there exists some $x_3 \in K \setminus N_{\varepsilon_0}(x_1) \cup N_{\varepsilon_0}(x_2)$ and in general

$$\exists x_n \in K \setminus (N_{\varepsilon_0}(x_1) \cup \dots \cup N_{\varepsilon_0}(x_{n-1}))$$

Therefore we have that for $n \neq m$

$$d(x_n, x_m) \geq \varepsilon_0$$

Therefore x_n doesn't have a convergent subsequence; contradiction.

□

Lemma 1.38. *Every subset of a totally bounded set is totally bounded.*

Proof. $K \subseteq X$ totally bounded, and $E \subseteq K$; take $\varepsilon > 0$, therefore $\exists x_1, x_2, \dots, x_t \in K$ s.t

$$K \subseteq N_{\varepsilon/2}(x_1) \cup \dots \cup N_{\varepsilon/2}(x_t)$$

Ignore the neighborhood that don't intersect E , renumbering we have x_1, \dots, x_l s.t

$$N_{\varepsilon/2}(x_i) \cap E \neq \emptyset$$

We get

$$E \subseteq N_{\varepsilon/2}(x_1) \cup \dots \cup N_{\varepsilon/2}(x_l)$$

Take $y_i \in N_{\varepsilon/2}(x_i) \cap E$, then

$$N_{\varepsilon}(y_i) \supseteq N_{\varepsilon/2}(x_i)$$

Therefore $E \subseteq N_{\varepsilon}(y_1) \cup \dots \cup N_{\varepsilon}(y_l)$ and hence E is totally bounded. \square

Example 1.39. $[0, 1]$ is totally bounded (since it is compact), therefore $(0, 1)$ is totally bounded.

Proposition 1.40. (X, d) Metric space, $K \subseteq X$ is complete and totally bounded then K is compact.

Proof. Assume K is not compact, fix $\mathcal{G} = \{G_{\alpha}\}_{\alpha \in I}$ an open cover of K with no finite subcover. Let $\varepsilon = 1$, K is totally bounded then K can be covered by finitely many 1-balls with center in K . Hence $\exists x_1 \in K$ s.t $K_1 = N_1(x_1) \cap K$ cannot be covered by finitely many open sets in \mathcal{G} . But K_1 is totally bounded (subset of K), now let $\varepsilon = 1/2$, $\exists x_2 \in K_1$ s.t $K_2 = N_{1/2}(x_2) \cap K_1$ cannot be covered by finitely many open sets in \mathcal{G} . Similarly, we construct a sequence x_n and

$$K_n = N_{1/n}(x_n) \cap K_{n-1} \quad x_n \in K_{n-1}$$

such that K_n cannot be covered by finitely many G_{α} 's. Notice that

$$K_0 = K \supseteq K_1 \supseteq K_2 \supseteq \dots$$

Claim. x_n is Cauchy. Indeed, let $\varepsilon > 0$ then $\exists N$ s.t

$$1/N < \varepsilon/2$$

Now take $m, n > N$, then

$$x_n \in K_{n-1} \subseteq K_N, \quad x_m \in K_{m-1} \subseteq K_N$$

therefore $x_n, x_m \in K_N = N_{1/N}(x_N) \cap K_{N-1}$, and hence

$$d(x_n, x_m) \leq 2/N < \varepsilon$$

By completeness of K , we get that x_n converges to some $x_0 \in K$; since $\mathcal{G} = \{G_{\alpha}\}$ is an open

cover of K , there exists $\alpha_0 \in I$ s.t $x_0 \in G_{\alpha_0}$ (open set). So $\exists \varepsilon_0$ s.t

$$N_{\varepsilon_0}(x_0) \subseteq G_{\alpha_0}$$

But $x_n \rightarrow x_0$ then $\exists N_1$ s.t

$$x_N \in N_{\varepsilon_0/2}(x_0) \text{ and } 1/N < \varepsilon_0/2$$

Therefore

$$K_N \subseteq N_{1/N}(x_n) \subseteq N_{\varepsilon_0}(x_0) \subseteq G_{\alpha_0}$$

contradiction. □

Theorem 1.41. V finite dimensional product normed space (over \mathbb{R}). Let $S = \{v_1, \dots, v_n\}$ a basis of V . Then the cube

$$Q_S = \{r_1v_1 + r_2v_2 + \dots + r_nv_n : 0 \leq r_i \leq 1\}$$

is compact.

Proof. Let $x^k = r_{1,k}v_1 + r_{2,k}v_2 + \dots + r_{n,k}v_n$ a sequence in Q_S , now clearly

$$r_{1,k} \text{ is a sequence in } [0, 1], \text{ compact set}$$

Therefore $r_{1,k}$ has a convergent subsequence r_{1,k_n} . Similarly $r_{2,k_l} \in [0, 1]$ therefore r_{2,k_l} has a convergent subsequence $r_{2,k_{l_t}}$. Doing this n times we get that \exists a subsequence $\{k_j\}$ s.t

$$r_{1,k_j}, r_{2,k_j}, \dots, r_{n,k_j} \rightarrow r_1, r_2, \dots, r_n$$

take

$$x_0 = r_1v_1 + \dots + r_nv_n$$

We can show that

$$\|x_0 - x^{k_j}\| \leq |r_{1,k_j} - r_1|\|v_1\| + \dots + |r_{n,k_j} - r_n|\|v_n\| \rightarrow 0$$

then

$$x^{k_j} \rightarrow 0$$

Hence Q_S is (sequentially) compact. □

Theorem 1.42. Let V be a finite inner product space then

$$K \subseteq V \text{ is compact} \iff K \text{ closed and bounded}$$

Proof. • K compact \implies complete \implies closed and \implies totally bounded \implies bounded.

- K is bounded, therefore $\exists R > 0$ s.t $K \subseteq B_{\|\cdot\|}(0, R)$. Let $S = \{v_1, \dots, v_n\}$ be a basis of V and Take $x \in K$ therefore

$$x = r_1v_1 + \dots + r_nv_n$$

$$\text{then } r_i = \|\langle x, v_i \rangle\| \leq \|x\| \cdot \|v_i\|$$

$$B_{\|\cdot\|}(0, R) \subseteq \{r_1 v_1 + \cdots + r_n v_n : r_i \leq R\}$$

Therefore $K \subseteq$ some cube and hence K is compact. □

Proposition 1.43. *In a finite dimensional inner product space we get that bounded \iff totally bounded.*

Proof. \implies Bounded \implies can be put in a cube (compact) \implies Totally bounded.
 \Leftarrow True in general. □

1.4 Continuity

Definition 1.44. X, Y metric spaces and $E \subseteq X$ and $f: E \rightarrow Y$ map; $x_0 \in E'$ we say that

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

iff for every $\varepsilon > 0 \exists \delta > 0$ s.t

$$f(x) \in N_\varepsilon(y_0) \text{ for } x \in N_\delta^*(x_0) \cap E$$

Definition 1.45. $f: E \rightarrow Y$ continuous at x_0 iff for every $\varepsilon > 0, \exists \delta > 0$ s.t $f(x) \in N_\varepsilon(y_0)$ for $x \in N_\delta(x_0) \cap E$; i.e

$$d_Y(f(x), f(x_0)) < \varepsilon \text{ for } 0 < d_X(x, x_0) < \delta$$

Remark 1.46. f is continuous at $x_0 \iff x_0$ is an isolated point or $x_0 \in E'$ and $\lim_{x \rightarrow x_0} f(x) = f(x_0)$.

Proposition 1.47. $\lim_{x \rightarrow x_0} f(x) = y_0$ iff for every sequence $x_n \rightarrow x_0$ we have that $f(x_n) \rightarrow y_0$. Therefore f is continuous at x_0 iff for every $x_n \rightarrow x_0$ we have that $f(x_n) \rightarrow f(x_0)$.

Corollary 1.48. if f, g are continuous then $f + g, cf, f \cdot g, f/g$ are all continuous at x_0 .

Example 1.49. $f(x, y) = \frac{xy}{x^2 + y^2}$ we want to find

$$\lim_{(x,y) \rightarrow (0,0)} f(x, y)$$

Take the sequence $(x_n, y_n) = (1/n, 0)$ then

$$\lim_{n \rightarrow \infty} (x_n, y_n) = (0, 0)$$

and

$$f(x_n, y_n) = 0 \rightarrow 0 \text{ as } n \rightarrow \infty$$

Now take $(x_n, y_n) = (1/n, 1/n)$, then

$$\lim_{n \rightarrow \infty} (x_n, y_n) = (0, 0)$$

but

$$f(x_n, y_n) = \frac{1/n \cdot 1/n}{1/n^2 + 1/n^2} = 1/2 \rightarrow 1/2 \text{ as } n \rightarrow \infty$$

Therefore

$$\lim_{n \rightarrow \infty} f(x, y) \text{ does not exist}$$

Now we do the same for

$$f(x, y) = \frac{x^2 y}{x^2 + y^2}$$

Notice that

$$0 \leq |f(x, y)| = \left| \frac{x^2}{x^2 + y^2} y \right| \leq |y|$$

but

$$\lim_{(x, y) \rightarrow (0, 0)} |y| = 0$$

therefore by squeeze we get that

$$\lim_{(x, y) \rightarrow 0} |f(x, y)| = 0$$

Hence

$$\lim_{(x, y) \rightarrow 0} f(x, y) = 0$$

Some useful inequalities

$$x^2 + y^2 \geq x^2$$

$$x^2 + y^2 \geq y^2$$

$$x^2 + y^2 \geq 2(xy)$$

Example 1.50. Let $(V, \|\cdot\|)$ a normed space and

$$\varphi: V \rightarrow \mathbb{R}, \quad \varphi(x) = \|x\|$$

Then φ is continuous.

Let x_n a sequence in V that converges to x_0 in $(V, \|\cdot\|)$ therefore

$$\|x_n - x_0\| \rightarrow 0$$

We want to show that $\varphi(x_n) \rightarrow \varphi(x_0)$ i.e that

$$|\|x_n\| - \|x_0\|| = 0$$

But as shown in PS1, we have that

$$|\|x_n\| - \|x_0\|| \leq \|x_n - x_0\|$$

Proposition 1.51. $f: X \rightarrow Y$ is continuous iff $f^{-1}(G)$ open in X for every G open in Y iff $f^{-1}(H)$ is closed in X for every H closed in Y .

Proof. \implies let x_n be a sequence in $f^{-1}(H)$ that converges to x_0 in X . Hence $f(x_n)$ is a sequence in H by continuity of f we have that

$$\lim_{n \rightarrow \infty} f(x_n) = f(x_0)$$

Moreover since H is closed, we get that $f(x_0) \in H$. Hence

$$x_0 \in f^{-1}(H)$$

Therefore if f is continuous $f^{-1}(G)$ is open in X for every G open in Y .

\Leftarrow Now we show that if $f^{-1}(G)$ is open in X for all G open in Y then f is continuous. Fix $x_0 \in X$, let $\varepsilon > 0$ then consider the open set in Y $N_\varepsilon(f(x_0))$ hence $f^{-1}(N_\varepsilon(f(x_0)))$ is open in X . But $x_0 \in f^{-1}(N_\varepsilon(f(x_0)))$ hence $\exists \delta > 0$ s.t

$$f(N_\delta(x_0)) \subseteq N_\varepsilon(f(x_0))$$

□

Example 1.52. $E = \{x, y \in \mathbb{R}^2 : y \geq x\}$ is open in \mathbb{R}^2 , define the continuous function

$$f: \mathbb{R}^2 \rightarrow \mathbb{R}, f(x, y) = y - x$$

Then

$$E = f^{-1}((0, \infty))$$

which is an open set.

Example 1.53. $f: X \rightarrow Y$ continuous and consider the c -level set with $c \in Y$

$$E = \{x \in X : f(x) = c\} = f^{-1}(\{c\})$$

is closed in X .

Therefore $E = \{(x, y) : x^2 + 3y^2 = 1\}$ is closed in \mathbb{R}^2 , S^n is closed.

Remark 1.54. V, W normed spaces and

$$\|\cdot\|_a, \|\cdot\|_b \text{ equivalent norms on } V$$

so $f: (V, \|\cdot\|_a) \rightarrow W$ is continuous iff $f: (V, \|\cdot\|_b) \rightarrow W$ is continuous.

Equivalently: V, W normed spaces and

$$\|\cdot\|_a, \|\cdot\|_b \text{ equivalent norms on } W$$

so $f: V \rightarrow (W, \|\cdot\|_a)$ is continuous iff $f: V \rightarrow (W, \|\cdot\|_b)$ is continuous.

Example 1.55. Take $f: \mathbb{R}^m \rightarrow \mathbb{R}^n$ defined as follows

$$f(x) = (f_1(x), \dots, f_n(x)) \quad x = (x_1, \dots, x_m)$$

Then f is continuous iff $f_i: \mathbb{R}^m \rightarrow \mathbb{R}$ is continuous for $1 \leq i \leq n$.

Theorem 1.56 (Extreme Value Theorem). X, Y metric spaces and $K \subseteq X$ compact and $f: K \rightarrow Y$ continuous. Then $f(K)$ is compact.

Corollary 1.57. If $Y = \mathbb{R}$, then the above implies that f attains its maximum and minimum in K .

Proof. Let y_n be a sequence in $f(K)$, therefore $\exists x_n \in K$ s.t $f(x_n) = y_n$. Since K is compact, x_n has a convergent subsequence x_{n_k} that converges to $x_0 \in K$, using continuity we get that

$$y_{n_k} \rightarrow f(x_0) \in f(K)$$

□

Theorem 1.58. All norms in a finite dimensional vector space are equivalent.

Proof. V is a finite dimensional vector space, then by PS1 we can define an inner product on V . Let $\|\cdot\|$ be the norm induced by our choice of inner product. Let $S = \{v_1, \dots, v_n\}$ an orthonormal basis of V . Let $\|\cdot\|_a$ be an arbitrary norm on V , we will show that $\|\cdot\|$ and $\|\cdot\|_a$ are equivalent. Let $x \in V$, then

$$x = r_1 v_1 + \dots + r_n v_n, \quad |r_i| = \langle x, v_i \rangle \leq \|x\|$$

Therefore $\forall x \in V$

$$\begin{aligned} \|x\|_a &= \|r_1 v_1 + \dots + r_n v_n\| \\ &\leq |r_1| \|v_1\|_a + \dots + |r_n| \|v_n\|_a \\ &\leq (\|v_1\|_a + \dots + \|v_n\|_a) \cdot \|x\| \\ &\leq c_1 \cdot \|x\| \end{aligned}$$

Now define the function $\varphi: (V, \|\cdot\|) \rightarrow \mathbb{R}$ s.t

$$\varphi(x) = \|x\|_a$$

We claim that φ is continuous, indeed let $x_n \in V$ s.t $x_n \rightarrow x_0$ in $\|\cdot\|$. Therefore

$$\begin{aligned} |\varphi(x_n) - \varphi(x_0)| &= |\|x_n\|_a - \|x_0\|_a| \\ &\leq \|x_n - x_0\| \\ &\leq c_1 \|x_n - x_0\| \rightarrow 0 \end{aligned}$$

Therefore $\lim_{n \rightarrow \infty} \varphi(x_n) = \varphi(x_0)$, so φ is continuous. Let $K = \{x \in V : \|x\| = 1\}$ closed and bounded in the inner product space $(V, \|\cdot\|)$. Then it is compact in $(V, \|\cdot\|)$ therefore $\varphi(K)$ is compact in \mathbb{R} so it attains its minimum in K . Notice that $\varphi(x) = 0 \iff x = 0$, hence $\forall x \in K, \varphi(x) > 0$. Then

$$\min_{x \in K} \varphi(x) > 0$$

Let $x \in V$ and $x \neq 0$ then $\frac{x}{\|x\|} \in K$, therefore

$$\varphi\left(\frac{x}{\|x\|}\right) = \left\|\frac{x}{\|x\|}\right\|_a \geq \min \varphi(K)$$

Therefore

$$\|x\|_a \geq \min(\varphi(K))\|x\|$$

□

Corollary 1.59. *If V is finite dimensional then compact iff closed and bounded (independently of the choice of norm).*

1.5 Connected sets

Definition 1.60. (X, d) metric space and $\Omega \subseteq X$, we say that Ω is disconnected iff $\exists U, V \subseteq X$ open sets s.t

$$\Omega = (U \cup V) \cap \Omega$$

With $U \cap \Omega, V \cap \Omega \neq \emptyset$ and $U \cap \Omega \cap V = \emptyset$.

Otherwise Ω is connected.

Example 1.61. Let $\Omega = (1, 2] \cup (3, 4)$ disconnected, since

$$\Omega \subseteq (0, 2.5) \cup (2.5, 17)$$

\mathbb{Q} is disconnected in \mathbb{R} since

$$\mathbb{Q} \subseteq (-\infty, \sqrt{2}) \cup (\sqrt{2}, \infty)$$

Proposition 1.62. *X is a connected metric space iff the only sets that are both open and closed are \emptyset, X .*

Proof. \implies Assume that U is open and closed and $U \neq \emptyset, X$ then U^c is also not open and closed and not in $\{\emptyset, X\}$ but

$$X = U \cup U^c$$

Therefore X is disconnected.

\Leftarrow Assume that X is disconnected, then $X = U \cup V$ with U, V open disjoint and nonempty. Therefore $V = U^c$ and U is open then V is closed. Hence V is a clopen set of X but $V \notin \{\emptyset, X\}$ contradiction.

□

Definition 1.63. $I \subseteq \mathbb{R}$ is an interval for $x, y \in I$ s.t $x < y$ then $z \in I, \forall x < z < y$.

Theorem 1.64. $\Omega \subseteq \mathbb{R}$ is connected iff Ω is an interval.

Proof. \Rightarrow Assume that Ω is not an interval, therefore $\exists x, y \in \Omega$ s.t $x < z < y$ and $z \notin \Omega$, therefore

$$\Omega \subseteq (-\infty, z) \cup (z, \infty)$$

Therefore Ω is disconnected.

\Leftarrow Let Ω be an interval. Suppose Ω is not connected, then $\exists U, V$ open and disjoint in Ω in \mathbb{R} s.t

$$\Omega \subseteq U \cup V$$

Then $\exists x \in U \cap \Omega$ and $y \in V \cap \Omega$ and assume wlog that $x < y$; therefore $[x, y] \subseteq \Omega$ consider

$$z = \sup \{[x, y] \cap U\}$$

Note that $z \in [x, y]$ therefore $z \in \Omega$ therefore either $z \in U$ or $z \in V$. Notice that $z > x$ since $x \in U$ an open set. Also notice that $z < y$ since if $z = y \in V$ we have that $\exists \varepsilon_1$ s.t

$$(y - \varepsilon, y) = (z - \varepsilon, z) \subseteq [x, y] \cap V$$

since z is the sup we get that

$$(z - \varepsilon, z) \cap [x, y] \cap U \neq \emptyset$$

contradiction.

Therefore $z \in (x, y)$, suppose that $z \in U$ then $(z, z + \varepsilon_0) \subseteq (x, y) \cap U$ contradiction because $z + \varepsilon_0/2 > z = \sup [x, y] \cap U$. Suppose that $z \in V$ then $\exists \varepsilon_1$ s.t $(z - \varepsilon_1, z) \subseteq (x, y) \cap V$. Since z is the sup then $\exists z^*$ an element s.t

$$z - \varepsilon_1 < z^* < z \text{ and } z^* \in U$$

$z^* \in U \cap V$ contradiction.

□

Theorem 1.65 (Intermediate Value Theorem). $f: X \rightarrow Y$ continuous, if X is connected then $f(X)$ is connected. In particular if $Y = \mathbb{R}$ we have the following:

- If $y_0 < y_1 \in f(X)$ then $\forall \alpha$ s.t $y_0 < \alpha < y_1$ $f^{-1}(\{\alpha\}) \neq \emptyset$.

Proof. Suppose that $f(X)$ is disconnected, then $\exists U, V \subseteq Y$ open sets that separate $f(X)$. There-

fore

$$\begin{aligned} X &= f^{-1}(f(X)) \\ &= f^{-1}(U \cap V) \cap f(X) \\ &= [f^{-1}(U) \cup f^{-1}(V)] \cap X \end{aligned}$$

But f is continuous therefore $f^{-1}(U)$, $f^{-1}(V)$ are open and disjoint. Now $U \cap f(X) \neq \emptyset$ therefore $f^{-1}(U) \neq \emptyset$, same for V ; then X is disconnected. \square

Definition 1.66. X is path connected iff for every $a, b \in X$ there exists a continuous function

$$\alpha: [0, 1] \rightarrow X$$

s.t $\alpha(0) = a$, $\alpha(1) = b$.

Remark 1.67. Let X be a vector space, the case where for every $a, b \in X$ the line segment $(1-t)a + tb \in X \forall t \in [0, 1]$ we say that X is convex.

Example 1.68. $B_R(x_0)$ is always convex and hence path connected in a vector space. Indeed, take $a, b \in B_R(x_0)$ consider

$$\begin{aligned} \|(1-t)a + tb - x_0\| &< (1-t)\|a - x_0\| + t\|b - x_0\| \\ &< (1-t)R + R = R \end{aligned}$$

Proposition 1.69. (X, d) metric space then X is path connected $\implies X$ is connected.

Proof. Let X be disconnected then we have U, V nonempty disjoint and open

$$X = U \cup V$$

Then $\exists a \in U$ and $b \in V$, since X is path connected $\exists \alpha: [0, 1] \rightarrow X$ s.t

$$\alpha(0) = a, \quad \alpha(1) = b$$

$[0, 1]$ is connected, therefore $\alpha([0, 1])$ is also connected but

$$[0, 1] = \alpha^{-1}(U) \cup \alpha^{-1}(V)$$

But α is continuous, therefore $\alpha^{-1}(U) \ni 0$ and $\alpha^{-1}(V) \ni 1$ which is a contradiction since $[0, 1]$ is connected. \square

Theorem 1.70. $(V, \|\cdot\|)$ vector space and $\Omega \subseteq V$ open then Ω is connected iff Ω is path connected.

Proof. \Leftarrow True in general.

\implies We will show that Ω is closed by showing that Ω is open. Take $z \in \Omega$ so there is no path joining z to x_0 . But Ω is open, so $\exists \delta > 0$ s.t

$$B_\delta(z) \subseteq \Omega$$

Now every point in $B_\delta(z)$ cannot be joined to x_0 ; Hence

$$B_\delta(z) \subseteq A^c$$

so A^c is open.

Therefore A open, closed and nonempty in Ω a connected space, hence $A = \Omega$.

□

Example 1.71 (Topological sine). $\Omega = \{(x, \sin(1/x)) \mid x > 0\} \cup \{(0, 0)\}$; let

$$A = \{(x, \sin 1/x) : x > 0\}$$

then A is connected (image of a continuous function), with

$$\overline{A} = A \cup (\{0\} \times [-1, 1])$$

Therefore $A \subseteq \Omega \subseteq \overline{A}$.

Suppose that Ω was connected, take $x_0 > 0$, suppose that $\exists \alpha: [0, 1] \rightarrow \Omega$ from $(0, 0)$ to $(x_0, \sin 1/x_0)$. let $t^* = \sup \{t \in [0, 1] \mid x(t) = 0\}$; therefore

$$x(t^*) = 0$$

with $x(t^*) < 1$, hence

$$\alpha(t^*) = (0, 0)$$

Take $\varepsilon = 1$, then by the IVT $\exists t^* < t_1 < t^* + 1$ and $t_1 \in [0, 1]$ s.t

$$x(t_1) = \frac{1}{(4n_1 + 1)\pi/2}$$

Take $\varepsilon = 1/2$, then $\exists t_2 \in [0, 1]$ s.t $t^* < t_2 < t^* + 1/2$ s.t

$$x(t_2) = \frac{1}{(4n_2 + 1)\pi/2}$$

We take a sequence $t_k \in [0, 1]$ s.t $t^* < t_k < t^* + 1/k$

$$x(t_k) = \frac{1}{(4n_k + 1)\pi/2}$$

hence

$$\alpha(t_k) = (\frac{1}{(4n_k + 1)\pi/2}, 1)$$

With $t_k \rightarrow t^*$, therefore

$$\alpha(t^*) = (0, 1)$$

contradicting the continuity of α .

2 Differentiability

Review linear transformations.

Proposition 2.1. *Let V and W normed spaces $T: V \rightarrow W$ linear transformation, then the following are all equivalent:*

- T is continuous on V .
- T is continuous on 0_V .
- $\exists M > 0$ s.t

$$\|Tx\|_W \leq M\|x\|_V$$

Proof. \implies 2 Trivial.

2 \implies 3 T is continuous at 0, hence for $\varepsilon = 1 \exists \delta > 0$ s.t

$$\|Tx - T0\|_W < 1 \quad \forall \|x - 0\|_V < \delta$$

Therefore

$$\|Tx\|_W < 1 \quad \forall \|x\|_V < \delta$$

Take any $x \neq 0$ then

$$\left\| \frac{x}{\|x\|_V} \cdot \delta/2 \right\|_V = \delta/2$$

Therefore

$$\left\| T \left(\frac{x}{\|x\|_V} \cdot \delta/2 \right) \right\|_W < 1$$

therefore

$$\|Tx\|_W < \frac{2}{\delta} \|x\|_V$$

3 \implies 1

$$\|Tx - Ty\|_W = \|T(x - y)\|_W \leq M\|x - y\|_V$$

□

Example 2.2 (The derivative as a linear transformation). Take $C^1([0, 1])$ the space of continuously differentiable functions, with the sup norm

$$\|f\|_\infty = \sup |f(x)| : x \in [0, 1]$$

and take $C([0, 1])$ to be the space of continuous functions with the sup norm. Take

$$\frac{d}{dx} : C^1([0, 1]) \rightarrow C([0, 1])$$

Note that this is a linear transformation.

Is it continuous? If it was then $\exists M$ s.t

$$\|f\|_\infty \leq M \|f\|_\infty \quad \forall f$$

take $f_n(x) = x^n$, then

$$\|f_n\|_\infty = 1$$

but $f'_n(x) = nx^{n-1}$ with

$$\|f'_n\|_\infty = n$$

Contradiction to the Archimedean property.

Theorem 2.3. Let V be finite dimensional, and $T: V \rightarrow W$ linear transformation then T is continuous.

Proof. Let $\alpha = \{v_1, \dots, v_n\}$ be a basis of V . Let $x \in V$, then $\exists r_1, \dots, r_n$ s.t

$$x = r_1v_1 + \dots + r_nv_n$$

Since all norms are equivalent in a finite dimensional vector space, wlog work with the inner product norm and make α an orthonormal basis.

$$\begin{aligned} \|Tx\|_W &= \|r_1Tv_1 + \dots + r_nTv_n\|_W \\ &\leq |r_1|\|Tv_1\|_W + \dots + |r_n|\|Tv_n\|_W \end{aligned}$$

With

$$|r_i| = |\langle x, v_i \rangle| \leq \|x\|$$

Therefore

$$\|Tx\|_W \leq \|x\|_V(\|Tv_1\| + \dots + \|Tv_n\|) = M\|x\|$$

□

Motivation. In one dimension, $f: \mathbb{R} \rightarrow \mathbb{R}$ and $x_0 \in \mathbb{R}$

$$\lim_{x \rightarrow x_0} \frac{f(x) - f(x_0)}{x - x_0}$$

if it exists we say that f is differentiable at x_0 and we define

$$f'(x_0) = \lim_{x \rightarrow x_0} \frac{f(x) - f(x_0)}{x - x_0}$$

In other words, we can say that

$$\begin{aligned}
 L = \lim_{x \rightarrow x_0} \frac{f(x) - f(x_0)}{x - x_0} &\iff \lim_{x \rightarrow x_0} \left(\frac{f(x) - f(x_0)}{x - x_0} - L \right) = 0 \\
 &\iff \lim_{x \rightarrow x_0} \left(\frac{f(x) - f(x_0) - L(x - x_0)}{x - x_0} \right) = 0 \\
 &\iff \exists L \in \mathbb{R} \text{ s.t } f(x) = f(x_0) + L(x - x_0) + o(|x - x_0|) \text{ with } \frac{o(t)}{|t|} \rightarrow 0 \\
 &\iff \exists L \in \mathbb{R} \text{ s.t } f(x) = f(x_0) + L(x - x_0) + o(|x - x_0|) \\
 &\quad \text{with } \frac{o(|x - x_0|)}{|x - x_0|} \rightarrow 0 \text{ as } x \rightarrow x_0
 \end{aligned}$$

Definition 2.4. V, W normed spaces and $U \subseteq V$ open set with $x_0 \in U$ we say that f is differentiable at x_0 iff there exists a continuous linear transformation $T \in \mathcal{L}(V, W)$ s.t

$$f(x) = f(x_0) + T(x - x_0) + o(\|x - x_0\|_V)$$

or

$$f(x_0 + h) = f(x_0) + T(h) + o(h)$$

with $\frac{o(\|h\|_V)}{\|h\|_V} \rightarrow 0$ as $h \rightarrow 0$.

Iff $\exists T \in \mathcal{L}(V, W)$ s.t

$$\lim_{h \rightarrow 0} \frac{f(x_0 + h) - f(x_0) - T(h)_W}{\|h\|_V} = 0$$

Remark 2.5. If such a T exists, then it is unique.

Proof. Suppose that $\exists T_1, T_2$ continuous linear transformations in $\mathcal{L}(V, W)$ s.t

$$f(x_0 + h) = f(x_0) + T_1(h) + o_1(h), \quad f(x_0 + h) = f(x_0) + T_2(h) + o_2(h)$$

Therefore

$$(T_1 - T_2)(h) = (o_1 - o_2)(h) = o_3(h)$$

take v s.t $\|v\| = 1$, therefore

$$\begin{aligned}
 (T_2 - T_1)(\lambda v) &= \lambda o_3(\|\lambda v\|) \\
 \lambda(T_2 - T_1)(v) &= \lambda o_3(|\lambda|) \\
 (T_2 - T_1)(v) &= \frac{o_3(|\lambda|)}{\lambda}
 \end{aligned}$$

Therefore as $\lambda \rightarrow 0$ we get that

$$(T_2 - T_1)(v) = 0 \quad \forall v \text{ s.t } \|v\| = 1$$

Therefore $T_2 = T_1$

□

Definition 2.6. If f is differentiable at x_0 , then we let $Df(x_0)$ to be the linear transformation in

$\mathcal{L}(V, W)$ s.t

$$f(x) = f(x_0) + Df(x_0)(x - x_0) + o(\|x - x_0\|) \iff \lim_{x \rightarrow x_0} \frac{\|f(x) - f(x_0) - Df(x_0)(x - x_0)\|_W}{\|x - x_0\|_V}$$

Notation. $\mathbb{R}^n = \{x = (x_1, \dots, x_n) : x_i \in \mathbb{R}\}$ as rows vectors, observe $f: \mathbb{R}^{1 \times n} \rightarrow \mathbb{R}^{m \times 1}$ differentiable at x_0 then we have that

$$f(x) = f(x_0) + Df(x_0)(x - x_0) + o(\|x - x_0\|)$$

Then

$$Df(x_0) \in \mathcal{L}(\mathbb{R}^m, \mathbb{R}^n)$$

Then $Df(x_0)$ can be represented by a matrix called the *Jacobian matrix* $[Df(x_0)]$ (an $m \times n$ matrix). Therefore

$$Df(x_0)(x - x_0) = [Df(x_0)](x - x_0)^t$$

Note that we take a row vector and output a column vector.

Example 2.7. In 1 variable we have that

$$[Df(x_0)] = [f'(x_0)]$$

Then

$$Df(x_0): t \rightarrow [f'(x_0)] \cdot t = f'(x_0) \cdot t$$

Find the derivative

$$f(x, y) = x^2 + y^2$$

we get that

$$\begin{aligned} f((x, y) + (h_1, h_2)) &= f(x_0 + h_1, y_0 + h_2) \\ &= (x_0 + h_1)^2 + (y_0 + h_2)^2 \\ &= x_0^2 + y_0^2 + 2x_0h_1 + 2y_0h_2 + h_1^2 + h_2^2 \\ &= f(x_0, y_0) + (2x_0, 2y_0)(h_1, h_2) + o(\|h\|) \end{aligned}$$

since $\frac{h_1^2 + h_2^2}{\sqrt{h_1^2 + h_2^2}} \rightarrow 0$ so we get

$$Df(x_0, y_0): (v_1, v_2) \rightarrow (2x_0, 2y_0) \cdot (v_1, v_2)$$

therefore we get

$$[Df(x_0)] = \begin{bmatrix} 2x_0 & 2y_0 \end{bmatrix}$$

Example 2.8. Take $T: V \rightarrow W$ be a continuous linear transformation we want to find $DT(x)$; but

$$T(x_0 + h) = T(x_0) + T(h) + 0$$

therefore

$$DT(x_0) = T$$

Proposition 2.9. $f: V \rightarrow W$ if f is differentiable at x_0 therefore f is continuous at x_0 .

Proof. Let $f(x) = f(x_0) + Df(x_0)(h) + o(\|x - x_0\|)$ but notice that

$$o(\|x - x_0\|) \rightarrow 0$$

but $Df(x_0)$ is continuous hence $\exists M$ s.t

$$\|Df(x_0)(x - x_0)\| \leq M\|x - x_0\| \rightarrow 0 \implies \lim_{x \rightarrow x_0} f(x) = f(x_0)$$

therefore f is continuous. □

2.1 Partial derivatives

Definition 2.10. $U \subseteq \mathbb{R}^n$ open and $f: U \rightarrow \mathbb{R}$, then

$$\frac{\partial f}{\partial x_i}(x) = \lim_{\substack{h \rightarrow 0 \\ h \in \mathbb{R}}} \frac{f(x + he_i) - f(x)}{h}$$

We call this the partial derivative w.r.t x_i .

Example 2.11.

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

We have shown that this is not continuous at $(0, 0)$ hence f is not differentiable at $(0, 0)$. But notice that

$$\frac{f(h, 0) - f(0, 0)}{h} = 0 \rightarrow 0 \text{ as } h \rightarrow 0$$

therefore $\frac{\partial f}{\partial x}(0, 0)$ exists and equal to 0. Similarly we get that $\frac{\partial f}{\partial y}(0, 0)$ exists and equal to 0.

What if $f: U \rightarrow \mathbb{R}$ is differentiable at x_0 then $Df(x_0) \in \mathcal{L}(R^n, \mathbb{R})$ hence $[Df(x_0)]$ is a $1 \times n$ matrix.

$$f(x_0 + he_1) = f(x_0) + Df(x_0)(he_1) + o(|h|)$$

therefore

$$\frac{f(x_0 + he_1) - f(x_0)}{h} = Df(x_0)e_1 + \frac{o(h)}{h} \rightarrow Df(x_0)e_1$$

therefore $\frac{\partial f}{\partial x_1}$ exists and is the first entry in $[Df(x_0)]$. Similarly we get $\frac{\partial f}{\partial x_i}$ exists and is the i 'th entry in $[Df(x_0)]$. We get that

$$[Df(x_0)] = \left(\frac{\partial f}{\partial x_1}, \dots, \frac{\partial f}{\partial x_n} \right)$$

this is called the gradient of f denoted ∇f .

If f is differentiable then

$$\begin{aligned} f(x) &= f(x_0) + \nabla f(x_0)(x - x_0)^t + o(|x - x_0|) \\ &= f(x_0) + \nabla f(x_0) \cdot (x - x_0) + o(|x - x_0|) \end{aligned}$$

Theorem 2.12. $U \subseteq \mathbb{R}^n$ open, and $f: U \rightarrow \mathbb{R}$ is differentiable iff all partial derivatives exist and

$$\lim_{h \rightarrow 0} \frac{\|f(x_0 + h) - f(x_0) - \nabla f(x_0) \cdot h\|}{|h|} = 0$$

Proof. \Rightarrow Suppose that f is differentiable, then

$$f(x_0 + h) = f(x_0) + Df(x_0)h + o(h)$$

Therefore

$$f(x_0 + he_i) = f(x_0) + Df(x_0)he_i + o(h)$$

Hence

$$\lim_{h \rightarrow 0} \frac{f(x_0 + he_i) - f(x_0)}{\|h\|} = Df(x_0)e_i$$

Therefore all partial derivatives exist with

$$\nabla f(x_0) \cdot e_i = Df(x_0)e_i$$

Hence

$$\lim_{h \rightarrow 0} \frac{\|f(x_0 + h) - f(x_0) - \nabla f(x_0) \cdot h\|}{|h|} = 0$$

\Leftarrow Suppose that all partial derivatives exist and

$$\lim_{h \rightarrow 0} \frac{\|f(x_0 + h) - f(x_0) - \nabla f(x_0) \cdot h\|}{|h|} = 0$$

We claim that

$$[Df(x_0)] = \nabla f(x_0)$$

Indeed, notice that by definition

$$f(x_0 + h) - f(x_0) - \nabla f(x_0) \cdot h = o(h)$$

Hence

$$f(x_0 + h) = f(x_0) + \nabla f(x_0) \cdot h + o(h)$$

□

Example 2.13. $f(x, y) = \begin{cases} \frac{x^2 y}{x^2 + y^2} & \text{if } (x, y) \neq (0, 0) \\ 0 & \text{o.w} \end{cases}$ then

$$f_x(0, 0) = 0, \quad f_y(0, 0) = 0$$

and

$$\frac{f(h_1, h_2) - f(0, 0) - \nabla f(0, 0) \cdot (h_1, h_2)}{\sqrt{h_1^2 + h_2^2}} = \frac{h_1^2 h_2}{(h_1^2 + h_2^2) \sqrt{h_1^2 + h_2^2}}$$

take $h_1 = 1/n, h_2 = 1/n$ we get that

$$\frac{1}{1\sqrt{2}} \not\rightarrow 0$$

hence this is not differentiable.

Example 2.14. $f(x, y) = x \cdot \sqrt{y}$ for $x \in \mathbb{R}, y \geq 0$ then

$$\frac{\partial f}{\partial x} = \sqrt{y} \quad x \in \mathbb{R}, y \geq 0$$

$$\frac{\partial f}{\partial y} = \frac{x}{2\sqrt{y}} \quad x \in \mathbb{R}, y > 0$$

But what happens to $\frac{\partial f}{\partial y}$ at $(x, 0)$?

$$\begin{aligned} \frac{\partial f}{\partial y}(x, 0) &= \lim_{h \rightarrow 0} \frac{f(x, h) - f(x, 0)}{h} \\ &= \lim_{h \rightarrow 0} \frac{x\sqrt{h}}{h} \\ &= \lim_{h \rightarrow 0} \frac{x}{\sqrt{h}} \end{aligned}$$

If $x \neq 0$ we have that

$$\lim_{h \rightarrow 0} \frac{x}{\sqrt{h}} \text{ doesn't exist}$$

hence $\frac{\partial f}{\partial y}$ doesn't exist at $(x, 0)$ for $x \neq 0$. But for $x = 0$ we have that

$$\frac{\partial f}{\partial y}(0, 0) = 0$$

Now we ask, is $f(x, y) = x\sqrt{y}$ differentiable at $(0, 0)$. Indeed

$$\nabla f(0, 0) = (f_x(0, 0), f_y(0, 0)) = (0, 0)$$

and

$$\begin{aligned} \frac{|f((0, 0) + (h_1, h_2)) - f(0, 0) - (0, 0) \cdot (h_1, h_2)|}{\|h\|} &= \frac{|h_1 \sqrt{h_2}|}{\sqrt{h_1^2 + h_2^2}} \\ &= \sqrt{\frac{h_1^2 |h_2|}{h_1^2 + h_2^2}} \\ &\leq \sqrt{|h_2|} \rightarrow 0 \end{aligned}$$

therefore f is differentiable at $(0, 0)$.

Theorem 2.15. $U \subseteq \mathbb{R}^n$ and $f: U \rightarrow \mathbb{R}^n$ with $f = (f_1, \dots, f_m)$ then f is differentiable iff

$f_1, \dots, f_n: U \rightarrow \mathbb{R}$ are all differentiable. Then

$$[Df(x_0)]_{m \times n} = \begin{pmatrix} \nabla f_1(x_0) \\ \nabla f_2(x_0) \\ \vdots \\ \nabla f_m(x_0) \end{pmatrix}$$

Therefore

$$Df(x_0) \begin{pmatrix} h_1 \\ h_2 \\ \vdots \\ h_n \end{pmatrix} = \begin{pmatrix} \nabla f_1(x_0) \cdot h \\ \nabla f_2(x_0) \cdot h \\ \vdots \\ \nabla f_m(x_0) \cdot h \end{pmatrix}$$

Example 2.16. $f(x, y) = (x^2y^2, x^2 + y^2, x^2 - y^2)$ then

$$[Df(x, y)] = \begin{pmatrix} 2xy^2 & 2x^2y \\ 2x & 2y \\ 2x & -2y \end{pmatrix}$$

Therefore

$$[Df(1, 2)] = \begin{pmatrix} 2 & 4 \\ 2 & 4 \\ 2 & -4 \end{pmatrix}$$

Hence

$$Df(1, 2): \begin{pmatrix} h_1 \\ h_2 \end{pmatrix} \rightarrow \begin{pmatrix} 8h_1 + 4h_2 \\ 2h_1 + 4h_2 \\ 2h_1 - 4h_2 \end{pmatrix}$$

Proof. • Suppose that f is differentiable, then

$$f(x_0 + h) = f(x_0) + Df(x_0)h + o(\|h\|)$$

Taking the dot with $e_i \in \mathbb{R}^m$ we get

$$f_i(x_0 + h) = f_i(x_0) + Df(x_0)(h) \cdot e_i + o(\|h\|)$$

Therefore f_i is differentiable with

$$Df_i(x_0): h \rightarrow (Df(x_0)(h)) \cdot e_i$$

• Suppose that each f_i is differentiable then

$$f_i(x_0 + h) = f_i(x_0) + Df_i(x_0)h + o_i(\|h\|)$$

Therefore

$$f(x_0 + h) = f(x_0) + (Df_1(x_0)h, \dots, Df_m(x_0)h) + (o_1(\|h\|), \dots, o_m(\|h\|))$$

Hence f is differentiable with

$$Df(x_0) = \begin{pmatrix} Df_1(x_0) \\ Df_2(x_0) \\ \vdots \\ Df_m(x_0) \end{pmatrix}$$

but $Df_i(x_0) = \nabla f_i(x_0)$. i.e

$$Df(x_0): h \rightarrow (\nabla f_i(x_0) \cdot h)_{1 \leq i \leq m}$$

□

Remark 2.17. f, g differentiable, then $f + \alpha g$ is differentiable, and

$$D(f + \alpha g) = Df + \alpha Dg$$

2.2 Chain Rule.

Theorem 2.18. Let $U \subseteq \mathbb{R}^n$ and $f: U \rightarrow \mathbb{R}^m$ and $V \subset \mathbb{R}^m$ open and $g: V \rightarrow \mathbb{R}$, take $x_0 \in U$ s.t $f(x_0) \in V$.

If f is differentiable at x_0 , and g is differentiable at $f(x_0)$. Then $h := g \circ f$ is differentiable at x_0 with

$$Dh(x_0) = Dg(f(x_0)) \circ Df(x_0)$$

In matrix form

$$[Dh(x_0)] = [Dg(f(x_0))][Df(x_0)]$$

Example 2.19. $h: \mathbb{R}^m \rightarrow \mathbb{R}$ with

$$h(u, v) = g(x(u, v), y(u, v))$$

Let $f(u, v) = (x(u, v), y(u, v))$ Therefore

$$[Dh(u, v)] = [Dg(x(u, v), y(u, v))][Df(u, v)]$$

$$\begin{pmatrix} \frac{\partial h}{\partial u} & \frac{\partial h}{\partial v} \end{pmatrix} = \begin{pmatrix} \frac{\partial g}{\partial x} & \frac{\partial g}{\partial y} \end{pmatrix} \begin{pmatrix} \frac{\partial x}{\partial u} & \frac{\partial x}{\partial v} \\ \frac{\partial y}{\partial u} & \frac{\partial y}{\partial v} \end{pmatrix}$$

Therefore we get that

$$\begin{pmatrix} \frac{\partial h}{\partial u} & \frac{\partial h}{\partial v} \end{pmatrix} = \left(\frac{\partial g}{\partial x} \frac{\partial x}{\partial u} + \frac{\partial g}{\partial y} \frac{\partial y}{\partial u} \quad \frac{\partial g}{\partial x} \frac{\partial x}{\partial v} + \frac{\partial g}{\partial y} \frac{\partial y}{\partial v} \right)$$

Example 2.20. $\gamma: [a, b] \rightarrow \mathbb{R}^n$ and $f: \mathbb{R}^n \rightarrow \mathbb{R}$ with

$$h(t) = f(\gamma(t))$$

Then

$$\begin{aligned}
 [Dh(t)] &= [Df(\gamma(t))][D\gamma(t)] \\
 [h'(t)] &= (\nabla f(\gamma(t))) \begin{pmatrix} \gamma'_1 \\ \gamma'_2 \\ \vdots \\ \gamma'_n \end{pmatrix} \\
 &= \nabla f(\gamma(t)) \cdot \gamma'(t)
 \end{aligned}$$

i.e $h: [a, b] \rightarrow \mathbb{R}$.

Consequences.

1. Mean value theorem for $\mathbb{R}^n \rightarrow \mathbb{R}$
2. $Df(x) = 0 \implies f$ is constant (in an open and connected domain).
3. If $\frac{\partial f_i}{\partial x_i}$ exist and are continuous at x_0 then f is differentiable at x_0 .

Proof. f is differentiable at x_0 then

$$f(x) = f(x_0) + Df(x_0)(x - x_0) + o(|x - x_0|) \quad (*)$$

and g is diff at $f(x_0)$, therefore

$$g(y) = g(f(x_0)) + Dg(f(x_0))(y - f(x_0)) + o(|y - f(x_0)|) \quad (**)$$

We have

$$\begin{aligned}
 h(x) &= g(f(x)) \\
 &= g(f(x_0)) + Dg(f(x_0))(f(x) - f(x_0)) + o(|f(x) - f(x_0)|) \\
 &= h(x_0) + Dg(f(x_0))(Df(x_0)(x - x_0) + o(|x - x_0|)) + o(|f(x) - f(x_0)|) \\
 &= h(x_0) + (Dg(f(x_0)) \circ Df(x_0))(x - x_0) + (Dg(f(x_0))(o(|x - x_0|)) + o(|f(x) - f(x_0)|))
 \end{aligned}$$

We will be done if we show that

$$R(x) = (Dg(f(x_0))(o(|x - x_0|)) + o(|f(x) - f(x_0)|)) = o(|x - x_0|)$$

but since $Dg(f(x_0))$ is a linear transformation

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

But using continuity we get as $x \rightarrow x_0$

$$Dg(f(x_0)) \left(\frac{o(|x - x_0|)}{|x - x_0|} \right) \rightarrow Dg(0) = 0$$

On the other hand, $Df(x_0)$ is a continuous linear transformation, then $\exists M$ s.t

$$\|Df(x_0)v\| \leq Mv$$

Let $\varepsilon > 0$, $\exists \delta > 0$ s.t

$$|o(|y - f(x_0)|)| < \frac{\varepsilon}{M+2}|y - f(x_0)|, \quad \forall |u - f(x_0)| < \delta$$

Now f is differentiable at x_0 , then it is continuous. Hence $\exists \gamma > 0$ s.t for $|x - x_0| < \gamma$

$$|f(x) - f(x_0)| < \delta$$

Therefore for $|x - x_0| < \gamma$ we have

$$o(|f(x) - f(x_0)|) < \frac{\varepsilon}{M+2}|f(x) - f(x_0)|$$

By (*) this is equal to $\frac{\varepsilon}{M+2}|Df(x_0)(x - x_0) + o(|x - x_0|)|$ therefore

$$o(|f(x) - f(x_0)|) < \frac{\varepsilon}{M+2}M(|x - x_0| + o(|x - x_0|))$$

by choosing γ even smaller

$$o(|x - x_0|) < |x - x_0|$$

Then for $|x - x_0| < \gamma$, we get

$$\begin{aligned} o(|f(x) - f(x_0)|) &< \frac{\varepsilon}{M+2}(M+1)|x - x_0| \\ &< \varepsilon|x - x_0| \end{aligned}$$

□

Recall. $\gamma: [a, b] \rightarrow U \subseteq \mathbb{R}^n$ and $f: U \rightarrow \mathbb{R}$ differentiable. Then

$$h(t) = f(\gamma(t)) \text{ is differentiable, } h: [a, b] \rightarrow \mathbb{R}$$

with

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

Theorem 2.21 (Mean Value theorem). $U \subseteq \mathbb{R}^n$ open and convex domain, $f: U \rightarrow \mathbb{R}$ differentiable and $x, y \in U$ then $\exists c$ on the line segment joining x to y s.t

$$f(y) - f(x) = \nabla f(c) \cdot (y - x)$$

Proof. Take $\gamma: [0, 1] \rightarrow \mathbb{R}^n$ s.t

$$\gamma(t) = (1 - t)x + ty$$

notice that $\gamma(1) = y \in U$, $\gamma(0) = x \in U$, therefore by convexity

$$\gamma([0, 1]) \subseteq U$$

define

$$h(t) = f(\gamma(t)), \quad h: [0, 1] \rightarrow \mathbb{R}$$

Then by MVT on \mathbb{R} , $\exists t_0 \in [0, 1]$ s.t

$$f(y) - f(x) = f(\gamma(1)) - f(\gamma(0)) = h(1) - h(0) = h'(t_0)(1 - 0)$$

But $h'(t_0)$ (by chain rule)

$$\nabla f(\gamma(t)) \cdot \gamma'(t) = f'(\gamma(t)) \cdot (y - x)$$

Therefore

$$f(y) - f(x) = f'(\gamma(t_0)) \cdot (y - x)$$

letting $c = \gamma(t_0)$ which belongs to the line segment, as desired. \square

Corollary 2.22. Let $U \subseteq \mathbb{R}^n$ open, connected domain. Moreover, let $f: U \rightarrow \mathbb{R}^m$ differentiable s.t

$$Df(x) = 0 \quad \forall x \in U$$

Then f is constant.

Proof. • **Case** $m = 1$. U is open and connected, therefore it must be path connected. Fix $x_0 \in U$ and $y \in U$, let $A = \{x \in U: f(x) = f(x_0)\}$, note that $A \neq \emptyset$ and A is closed since

$$A = f^{-1}(\{f(x_0)\})$$

let $x \in A$, U is open then $\exists \delta > 0$ s.t

$$B(x, \delta) \subseteq U$$

but $B(x, \delta)$ is convex, then by the MVT for every $y \in B(x, \delta)$ $\exists c$ in the line joining x to y s.t

$$f(y) - f(x) = \nabla f(c) \cdot (y - x) = 0$$

Therefore

$$f(y) = f(x) = f(x_0) \quad \forall y \in B(x, \delta)$$

Therefore

$$B(x, \delta) \subseteq A$$

hence A is open. By connectedness of U we get that $A = U$.

- **General case.** Let $f = (f_1, \dots, f_m)$ therefore

$$[Df(x)] = \begin{pmatrix} \nabla f_1 \\ \vdots \\ \nabla f_m \end{pmatrix}$$

but $\nabla f_i = 0 \implies f_i = c_i$, therefore $f = c$.

□

Theorem 2.23. $U \subseteq \mathbb{R}^n$ and $f: U \rightarrow \mathbb{R}$ and fix $a = (a_1, \dots, a_n) \in U$ If $\frac{\partial f}{\partial x_i}$ exist in a neighborhood of a and $\frac{\partial f}{\partial x_i}$ are continuous at a . Then f is differentiable at a .

Corollary 2.24. If $f: U \rightarrow \mathbb{R}^m$ and $a \in U$ if $\frac{\partial f_i}{\partial x_j}$ exist in a neighborhood of a and are continuous at a , then f is differentiable at a .

Example 2.25. $f(x, y) = x\sqrt{y}$ is differentiable for $x \in \mathbb{R}$ and $y > 0$ and $f_x = \sqrt{y}$, $f_y = \frac{x}{2\sqrt{y}}$ which both exist and are continuous in a neighborhood of (x_0, y_0) for $x_0 \in \mathbb{R}$ and $y_0 > 0$. So f is differentiable at these points.

Proof. Let $B(a, r) \subset U$ and $|h| < R$ with $h = (h_1, \dots, h_n)$, then

$$\begin{aligned} f(a+h) - f(a) &= f(a+h_1e_1) - f(a) \\ &\quad + f(a+h_1e_1+h_2e_2) - f(a+h_1e_1) \\ &\quad - f(a+h_1e_1+h_2e_2) - f(a) \\ &\quad \vdots \\ &\quad + f(a+h) - f(a+h_1e_1+h_2e_2+\dots+h_{n-1}e_{n-1}) \\ &= \frac{\partial f}{\partial x_1}(a+c_1e_1) \cdot h_1 \\ &\quad \frac{\partial f}{\partial x_2}(a+h_1e_1+c_2e_2) \cdot h_2 \\ &\quad \vdots \\ &\quad + \frac{\partial f}{\partial x_n}(a+h_1e_1+\dots+h_{n-1}e_{n-1}+c_ne_n)h_n \end{aligned}$$

Where $c_i \in [0, h_i]$, therefore

$$\begin{aligned} f(a+h) - f(a) &= \sum_{i=1}^n \frac{\partial f}{\partial x_i}(a+h_1e_1+\dots+h_{i-1}e_{i-1}+c_ie_i)h_i \\ &= \frac{|f(a+h) - f(a) - \nabla f(a) \cdot h|}{h} \\ &= \frac{1}{|h|} \sum_{i=1}^n \left(\frac{\partial f}{\partial x_i}(a+h_1e_1+\dots+h_{i-1}e_{i-1}+c_ie_i) - \frac{\partial f}{\partial x_i}(a) \right) h_i \\ &\leq \sum_{i=1}^n \sum_{i=1}^n \left| \frac{\partial f}{\partial x_i}(a+h_1e_1+\dots+h_{i-1}e_{i-1}+c_ie_i) - \frac{\partial f}{\partial x_i}(a) \right| \frac{|h_i|}{|h|} \\ &\leq \sum_{i=1}^n \sum_{i=1}^n \left| \frac{\partial f}{\partial x_i}(a+h_1e_1+\dots+h_{i-1}e_{i-1}+c_ie_i) - \frac{\partial f}{\partial x_i}(a) \right| \end{aligned}$$

Let $h \rightarrow 0$ then $a + h_1 e_1 + \cdots + h_{i-1} e_{i-1} + c_i e_i \rightarrow a$. and use continuity to get

$$\lim_{h \rightarrow 0} \frac{|f(a+h) - f(a) - \nabla f(a) \cdot h|}{|h|} = 0$$

so f is differentiable at a . □

Remark 2.26. The converse is not true, since $f(x, y) = x\sqrt{y}$ is differentiable at $(0, 0)$ but f_y doesn't exist at $(x, 0) \forall x \neq 0$.

Corollary 2.27. $U \subset \mathbb{R}^n$ and $f: U \rightarrow \mathbb{R}^m$ differentiable,

$$Df: U \rightarrow \mathcal{L}(\mathbb{R}^n, \mathbb{R}^m)$$

Then Df is continuous on U iff all partial derivatives $\frac{\partial f_i}{\partial x_i}$ of f are continuous on U .

Proof. $Df(u) = \begin{pmatrix} \nabla f_1 \\ \nabla f_2 \\ \vdots \\ \nabla f_n \end{pmatrix}$ which is continuous iff every entry is continuous. □

Definition 2.28. V, W finite dimensional vector space $U \subset V$ open and $f: U \rightarrow W$ we say that f is C^1 iff f is differentiable and Df is continuous.

Note that this is equivalent to saying that all partial derivatives exist and are continuous.

2.3 Higher order derivatives.

Motivation. Let $\alpha = \{e_1, \dots, e_n\}$ basis of \mathbb{R}^n and $\hat{\alpha} = \{\hat{e}_1, \dots, \hat{e}_m\}$ basis of \mathbb{R}^m . Take

$$f: U \rightarrow \mathbb{R}^m, \quad \text{differentiable}$$

and

$$Df: U \rightarrow \mathcal{L}(\mathbb{R}^n, \mathbb{R}^m)$$

Define

$$\gamma = \{T_{11}, \dots, T_{1m}, T_{21}, \dots, T_{nm}\}$$

with

$$T_{i_0 j_0}: e_{i_0} \rightarrow \hat{e}_{j_0}, \quad e_{i \neq i_0} \rightarrow 0$$

Take $x \in U$ then

$$Df(x) \in \mathcal{L}(\mathbb{R}^n, \mathbb{R}^m)$$

with

$$Df(x) = r_{11}T_{11} + r_{12}T_{12} + \cdots + r_{nm}T_{nm}$$

Then

$$Df(x)e_j = r_{j1}\hat{e}_1 + \cdots + r_{jm}\hat{e}_m$$

therefore

$$[Df(x)] = \begin{pmatrix} \nabla f_1 \\ \vdots \\ \nabla f_m \end{pmatrix}$$

Therefore

$$Df(x)e_i = \begin{pmatrix} \frac{\partial f_1}{\partial x_i} \\ \vdots \\ \frac{\partial f_m}{\partial x_i} \end{pmatrix}$$

So we get that

$$r_{ij} = \frac{\partial f_j}{\partial x_i}$$

Therefore

$$[Df]_\gamma = \begin{pmatrix} \frac{\partial f_1}{\partial x_1} \\ \frac{\partial f_1}{\partial x_2} \\ \vdots \\ \frac{\partial f_m}{\partial x_n} \end{pmatrix}$$

Hence we get that

$$D[Df] = \begin{pmatrix} \nabla(f_1)_{x_1} \\ \nabla(f_1)_{x_2} \\ \vdots \\ \nabla(f_m)_{x_m} \end{pmatrix}$$

Note that

$$Df: U \rightarrow \mathcal{L}(\mathbb{R}^n, \mathbb{R}^m)$$

Hence

$$D(Df): U \rightarrow \mathcal{L}(\mathbb{R}^n, \mathcal{L}(\mathbb{R}^n, \mathbb{R}^m))$$

Therefore $D[Df]$ is a $mn \times n$ matrix, which we call the Hessian matrix.

Example 2.29. $f(x, y) = (x^2y^3, x^2 + y^3, x^3 + y^2)$ we have that

We get that

$$D^2(f) = \begin{pmatrix} (f_1)_{xx} & (f_1)_{xy} \\ (f_1)_{yx} & (f_1)_{yy} \\ (f_2)_{xx} & (f_2)_{xy} \\ (f_2)_{yx} & (f_2)_{yy} \\ (f_3)_{xx} & (f_3)_{xy} \\ (f_3)_{yx} & (f_3)_{yy} \end{pmatrix}$$

Remark 2.30. In the particular case where $m = 1$ we have that $f: U \rightarrow \mathbb{R}$ therefore

$$Df = \begin{pmatrix} \nabla f_{x_1} \\ \nabla f_{x_2} \\ \vdots \\ \nabla f_{x_n} \end{pmatrix}$$

In the $n = 2$ case we get

$$\begin{pmatrix} \frac{\partial}{\partial x} \frac{\partial f}{\partial x} & \frac{\partial}{\partial y} \frac{\partial f}{\partial x} \\ \frac{\partial}{\partial x} \frac{\partial f}{\partial y} & \frac{\partial}{\partial y} \frac{\partial f}{\partial y} \end{pmatrix}$$

Theorem 2.31 (Mixed derivatives theorem). $U \subseteq \mathbb{R}^n$ and $f: U \rightarrow \mathbb{R}$, take $x_0 \in U$. If $\frac{\partial f}{\partial x_i \partial x_j}$ and $\frac{\partial f}{\partial x_j \partial x_i}$ exist in a neighborhood of x_0 and are continuous at x_0 then they are equal.

Proof. $n = 2$, assume that $\frac{\partial f}{\partial x_i \partial x_j}$ and $\frac{\partial f}{\partial x_j \partial x_i}$ exist in a neighborhood of a and are continuous at $a = (a_1, a_2)$. Define for $h, k > 0$

$$w(h, k) = f(a_1 + h) - f(a_1 + h, a_2) - f(a_1, a_2 + h) + f(a_1, a_2)$$

Define

$$v(h, k) = f(a_1 + h, a_2 + k) - f(a_1 + h, a_2)$$

and

$$u(h, k) = f(a_1 + h, a_2 + k) - f(a_1, a_2 + k)$$

Notice that by MVT

$$\begin{aligned} w(h, k) &= v(h, k) - v(0, k) \\ &= v_h(c, k) \quad c \in [0, h] \\ &= (f_{x_1}(a_1 + c, a_2 + k) - f_{x_1}(a_1 + c, a_2)) h \\ &= (f_{x_1 x_2}(a_1 + c, a_2 + d)) kh \quad d \in [0, k] \end{aligned}$$

Similarly

$$\begin{aligned} w(h, k) &= u(h, k) - u(h, 0) \\ &= u_k(h, d') k \quad d' \in [0, k] \\ &= (f_{x_2}(a_1 + h, a_2 + d') - f_{x_2}(a_1, a_2 + d')) k \\ &= (f_{x_2 x_1}(a_1 + c', a_2 + d')) kh \quad c' \in [0, h] \end{aligned}$$

Hence

$$f_{x_1 x_2}(a_1 + c, a_2 + d') = f_{x_2 x_1}(a_1 + c', a_2 + d')$$

then let $h, k \rightarrow 0$ by continuity we get

$$f_{x_1 x_2}(a_1, a_2) = f_{x_2 x_1}(a_1, a_2)$$

□

Definition 2.32. We say that $f \in C^k(U)$ ($k \in \mathbb{N}$) iff all partial derivatives of order k of f (of its components exist) and are continuous.

Taylor's theorem. In one variable, if g is C^1 then

$$g(t_0 + h) = g(t_0) + g'(t_0)h + o(|h|).$$

If g is C^2 , then

$$g(t_0 + h) = g(t_0) + g'(t_0)h + \frac{g''(t_0)}{2!}h^2 + o(|h|^2).$$

More generally, if g is C^k , then

$$g(t_0 + h) = \sum_{i=0}^k \frac{g^{(i)}(t_0)}{i!} h^i + o(|h|^k).$$

Now take $U \subset \mathbb{R}^n$ open and a function

$$f : U \rightarrow \mathbb{R}.$$

Fix $x_0 \in U$ and assume f is C^1 in U . Take \hat{u} a unit vector in \mathbb{R}^n , and define

$$g(t) = f(x_0 + t\hat{u}),$$

which is defined for $|t| \leq \delta$ small enough. Therefore

$$g \in C^1([-\delta, \delta]).$$

Hence

$$g(t) = g(0) + \frac{g'(0)}{1!} t + o(|t|).$$

But

$$g'(t) = \nabla f(x_0 + t\hat{u}) \cdot \hat{u},$$

so

$$g'(0) = \nabla f(x_0) \cdot \hat{u}.$$

Therefore

$$f(x_0 + t\hat{u}) = f(x_0) + (\nabla f(x_0) \cdot \hat{u}) t + o(|t|).$$

Now for $h \in \mathbb{R}^n$ small, write $h = |h|\hat{u}$ where $\hat{u} = \frac{h}{|h|}$ (for $h \neq 0$). Taking $t = |h|$ above gives

$$\begin{aligned} f(x_0 + h) &= f\left(x_0 + |h|\frac{h}{|h|}\right) \\ &= f(x_0) + \left(\nabla f(x_0) \cdot \frac{h}{|h|}\right) |h| + o(|h|) \\ &= f(x_0) + \nabla f(x_0) \cdot h + o(|h|) \\ &= f(x_0) + \sum_{i=1}^n f_{x_i}(x_0) h_i + o(|h|). \end{aligned}$$

Assume now that f is C^2 . Then g is C^2 and

$$g(t) = g(0) + g'(0)t + \frac{g''(0)}{2}t^2 + o(|t|^2).$$

We have

$$g'(t) = \sum_{i=1}^n \frac{\partial f}{\partial x_i}(x_0 + t\hat{u}) \hat{u}_i,$$

hence differentiating once more,

$$\begin{aligned} g''(t) &= \sum_{i=1}^n \left(\frac{\partial f}{\partial x_i}(x_0 + t\hat{u}) \right)' \hat{u}_i \\ &= \sum_{i=1}^n \left(\nabla \left(\frac{\partial f}{\partial x_i} \right) (x_0 + t\hat{u}) \cdot \hat{u} \right) \hat{u}_i \\ &= \sum_{i=1}^n \sum_{j=1}^n \frac{\partial^2 f}{\partial x_j \partial x_i}(x_0 + t\hat{u}) \hat{u}_j \hat{u}_i. \end{aligned}$$

Therefore

$$g''(0) = \sum_{i=1}^n \sum_{j=1}^n \frac{\partial^2 f}{\partial x_j \partial x_i}(x_0) \hat{u}_j \hat{u}_i.$$

Let $D^2f(x_0)$ denote the Hessian matrix

$$D^2f(x_0) = \left(\frac{\partial^2 f}{\partial x_i \partial x_j}(x_0) \right)_{1 \leq i, j \leq n}.$$

Then the previous expression can be written as the quadratic form

$$g''(0) = \langle D^2f(x_0)\hat{u}, \hat{u} \rangle.$$

Hence

$$g(t) = g(0) + (\nabla f(x_0) \cdot \hat{u})t + \frac{1}{2} \langle D^2f(x_0)\hat{u}, \hat{u} \rangle t^2 + o(|t|^2).$$

Equivalently, for $h \in \mathbb{R}^n$ small (again write $h = |h|\hat{u}$),

$$f(x_0 + h) = f(x_0) + \nabla f(x_0) \cdot h + \frac{1}{2} \langle D^2f(x_0)h, h \rangle + o(|h|^2).$$

Example ($n = 2$). If $f = f(x_1, x_2)$ and $h = (h_1, h_2)$, then

$$\begin{aligned} f(x_0 + h) &= f(x_0) + f_{x_1}(x_0)h_1 + f_{x_2}(x_0)h_2 \\ &\quad + \frac{1}{2} \begin{pmatrix} f_{x_1x_1}(x_0) & f_{x_1x_2}(x_0) \\ f_{x_2x_1}(x_0) & f_{x_2x_2}(x_0) \end{pmatrix} \begin{pmatrix} h_1 \\ h_2 \end{pmatrix} \cdot \begin{pmatrix} h_1 \\ h_2 \end{pmatrix} + o(|h|^2) \\ &= f(x_0) + f_{x_1}(x_0)h_1 + f_{x_2}(x_0)h_2 \\ &\quad + \frac{1}{2} \left(f_{x_1x_1}(x_0)h_1^2 + (f_{x_1x_2}(x_0) + f_{x_2x_1}(x_0))h_1h_2 + f_{x_2x_2}(x_0)h_2^2 \right) + o(|h|^2). \end{aligned}$$

(If $f \in C^2$, then $f_{x_1x_2}(x_0) = f_{x_2x_1}(x_0)$.)

This matches the pattern

$$\frac{f(x_0)}{0!} + \left(\frac{f_{x_1}(x_0)h_1}{1!} + \frac{f_{x_2}(x_0)h_2}{1!} \right) + \left(\frac{f_{x_1x_1}(x_0)h_1^2}{2!} + \frac{f_{x_1x_2}(x_0)h_1h_2}{1!1!} + \frac{f_{x_2x_2}(x_0)h_2^2}{2!} \right) + o(|h|^2).$$

Multi-index notation. Take $\alpha = (\alpha_1, \dots, \alpha_n)$ where $\alpha_i \in \mathbb{N} \cup \{0\}$. Define the order of α by

$$|\alpha| := \alpha_1 + \dots + \alpha_n,$$

and the factorial by

$$\alpha! := \alpha_1! \cdots \alpha_n!.$$

For $f \in C^{|\alpha|}$, define the α -partial derivative

$$D^\alpha f := \frac{\partial^{\alpha_1}}{\partial x_1^{\alpha_1}} \frac{\partial^{\alpha_2}}{\partial x_2^{\alpha_2}} \cdots \frac{\partial^{\alpha_n}}{\partial x_n^{\alpha_n}} f.$$

For $x \in \mathbb{R}^n$, define

$$x^\alpha := x_1^{\alpha_1} \cdots x_n^{\alpha_n}.$$

Example 2.33. Let $f(x, y, z) = x^2y^2z^3$. Then

$$\begin{aligned} \frac{D^{(1,2,1)}f}{(1,2,1)!} &= \frac{1}{1!2!1!} \frac{\partial}{\partial x} \frac{\partial^2}{\partial y^2} \frac{\partial}{\partial z} (x^2y^2z^3) \\ &= \frac{1}{2} \frac{\partial}{\partial x} \frac{\partial^2}{\partial y^2} (3x^2y^2z^2) \\ &= \frac{1}{2} \frac{\partial}{\partial x} (6x^2z^2) \\ &= 6xz^2. \end{aligned}$$

Theorem 2.34 (Taylor's theorem). Let $U \subset \mathbb{R}^n$ be open, $x_0 \in U$, and $f : U \rightarrow \mathbb{R}$ be C^k . Then

$$f(x_0 + h) = \sum_{0 \leq |\alpha| \leq k} \frac{D^\alpha f(x_0)}{\alpha!} h^\alpha + o(|h|^k) \quad (h \rightarrow 0),$$

equivalently,

$$f(x) = \sum_{0 \leq |\alpha| \leq k} \frac{D^\alpha f(x_0)}{\alpha!} (x - x_0)^\alpha + o(|x - x_0|^k) \quad (x \rightarrow x_0).$$

2.4 Extremum

Definition 2.35. $f : U \rightarrow \mathbb{R}$ $U \subseteq \mathbb{R}^n$ and $x_0 \in U$. We say that f has a local minimum at x_0 iff $\exists \delta > 0$ s.t $f(x) \geq f(x_0) \forall x \in B(x_0, \delta)$.

We say that f has a local maximum at x_0 iff $\exists \delta > 0$ s.t $f(x) \leq f(x_0) \forall x \in B(x_0, \delta)$

Theorem 2.36. If f has a local min (or max) at x_0 and f is differentiable at x_0 then $\nabla f(x_0) = 0$.

Proof. f has a local minimum at x_0 , then $\exists \delta > 0$ s.t $f(x) > f(x_0) \forall x \in B(x_0, \delta)$. Let $|t| < \delta$ and \hat{u} a unit vector; since f is differentiable at x_0 , with $x_0 + t\hat{u} \in B(x_0, \delta)$. Hence

$$f(x_0 + t\hat{u}) = f(x_0) + (\nabla f(x_0) \cdot \hat{u})t + o(|t|) \geq f(x_0)$$

Therefore

$$(\nabla f(x_0) \cdot \hat{u})t + o(|t|) \geq 0$$

- if $t \geq 0$ we have that

$$\nabla f(x_0) \cdot \hat{u} + \frac{o(|t|)}{t} \geq 0$$

let $t \rightarrow 0$

$$\nabla f(x_0) \cdot \hat{u} \geq 0$$

- if $t \leq 0$ we have that

$$\nabla f(x_0) \cdot \hat{u} + \frac{o(|t|)}{t} \leq 0$$

Let $t \rightarrow 0$ we have

$$\nabla f(x_0) \cdot \hat{u} \leq 0$$

Therefore $\forall \hat{u}, |\hat{u}| = 1$ we have that

$$\nabla f(x_0) \cdot \hat{u} = 0 \implies \nabla f(x_0) = 0$$

□

Example 2.37. Find local min/max of the following:

$$f(x, y) = x^2 + y^2$$

We have

$$\nabla f(x, y) = (2x, 2y) = (0, 0)$$

Hence

$$\begin{cases} 2x = 0 \\ 2y = 0 \end{cases} \implies (x, y) = 0$$

Therefore

$$f(0, 0) = 0 \leq x^2 + y^2 = f(x, y) \forall (x, y)$$

Therefore at $(0, 0)$ we have a minimum.

$$f(x, y) = x^2 - y^2$$

Therefore

$$\nabla f(x, y) = (2x, -2y) = 0 \implies x = y = 0$$

Notice that $f(0, 0) = 0$ Take $\delta > 0$, $B(0, \delta)$ take

$$f(\delta/2, 0) = \delta^2/4 > 0, \quad f(0, \delta) = -\delta^2/4 < 0$$

Therefore $(0, 0)$ is neither a local min nor a local max (it is a saddle point).

Second derivative test. Review the spectral theorem and diagonalization.

Suppose that A is symmetric and let $\lambda_1, \dots, \lambda_n$ be the eigenvalues of A with corresponding orthonormal basis of eigenvectors v_1, \dots, v_n , i.e

$$Av_i = \lambda_i v_i, \quad \langle v_i, v_j \rangle = \delta_{ij}$$

Let $v \in \mathbb{R}^n$, then we can write v as $r_1 v_1 + \dots + r_n v_n$, therefore

$$Av = (\lambda_1 r_1) v_1 + \dots + (\lambda_n r_n) v_n$$

Therefore we get that

$$\begin{aligned} \langle Av, v \rangle &= \langle r_1 \lambda_1 v_1 + \dots + r_n \lambda_n v_n, r_1 v_1 + \dots + r_n v_n \rangle \\ &= r_1^2 \lambda_1 + r_2^2 \lambda_2 + \dots + r_n^2 \lambda_n \end{aligned}$$

Definition 2.38. Let A be an $n \times n$ symmetric matrix we say that A is positive definite iff

$$\langle Av, v \rangle > 0 \quad \forall v \neq 0$$

iff all eigenvalues of A are positive.

We say that it is positive semi-definite iff

$$\langle Av, v \rangle \geq 0 \quad \forall v \neq 0$$

iff all eigenvalues of A are nonnegative.

We define negative (semi) definite similarly.

Otherwise, we say that A is indefinite (at least two eigenvalues of opposite signs).

Remark 2.39. A is symmetric, then $\lambda_i \geq 0$. Wlog $\lambda_1 \leq \lambda_2 \leq \dots \leq \lambda_n$, therefore

$$\lambda_1 |v|^2 \leq \langle Av, v \rangle \leq \lambda_n |v|^2$$

A is positive semi-definite $\iff \exists c \geq 0$ s.t

$$\langle Av, v \rangle \geq c |v|^2$$

Definition 2.40. $U \subset \mathbb{R}^n$ and $f: U \rightarrow \mathbb{R}$ a C^1 function. We say that x_0 is a critical point of f iff

$$\nabla f(x_0) = 0$$

(Candidate of extremum)

We say that $f \in C(U)$ attains a saddle point at x_0 iff $\nabla f(x_0) = 0$ but f doesn't attain a local extremum at x_0 .

Theorem 2.41 (Second derivative test). $U \subset \mathbb{R}^n$ open and $f: U \rightarrow \mathbb{R}$ a C^2 function, take $x_0 \in U$ s.t

$$\nabla f(x_0) = 0$$

1. If $D^2 f(x_0)$ is positive definite, then f attains a local minimum.
2. If $D^2 f(x_0)$ is negative definite, then f attains a local maximum.
3. If $D^2 f(x_0)$ has two eigenvalues with opposite signs, then f has a saddle point at x_0 .

The test is inconclusive if $Df(x_0)$ is semi-definite.

Proof. By Taylor's theorem, we have that

$$f(x) = f(x_0) + \nabla f(x_0) \cdot (x - x_0) + \frac{1}{2} \langle D^2 f(x_0)(x - x_0), x - x_0 \rangle + o(|x - x_0|^2)$$

(Since $\nabla f(x_0) = 0$) we get

$$\begin{aligned} f(x) - f(x_0) &= 1/2 \langle D^2 f(x_0)(x - x_0), x - x_0 \rangle + o(|x - x_0|^2) \\ &\geq C|x - x_0|^2 + o(|x - x_0|^2) && C > 0 \\ &= |x - x_0|^2 \left(C + \frac{o(|x - x_0|^2)}{|x - x_0|^2} \right) \\ &\geq 0 && \text{for } |x - x_0| < \delta \end{aligned}$$

Therefore

$$f(x) \geq f(x_0) \text{ for } |x - x_0| < \delta$$

Hence x_0 is a local minimum.

We can prove 2 the exact same way.

For 3, suppose that λ, λ' are two eigenvalues of $D^2 f(x_0)$ with

$$\lambda > 0, \quad \lambda' < 0$$

Let v, v' be the corresponding normalized eigenvectors. Take

$$x = x_0 + tv$$

Therefore

$$\begin{aligned}
 f(x_0 + tv) - f(x_0) &= \frac{1}{2} \langle D^2 f(x_0) tv, tv \rangle + o(|t|^2) \\
 &= \frac{1}{2} t^2 \left(\langle D^2 f(x_0) v, v \rangle + \frac{o(|t|^2)}{t^2} \right) \\
 &= \frac{1}{2} t^2 (\lambda + o(|t|^2)/t^2) \\
 &\geq 0
 \end{aligned}$$

t small enough

Similarly, for t small enough

$$f(x_0 + tv') - f(x_0) \leq 0$$

Hence f has a saddle point at x_0 . □

Example 2.42. Math 201:

- Solve $\nabla f(x_0) = 0$
- $D^2 f(x_0)$ find its eigenvalues.
- Apply second derivative test (if applicable).

2.5 Inverse function theorem.

Lemma 2.43 (Banach fixed point theorem). X complete metric space, $T: X \rightarrow X$ a contraction, i.e $\exists c \in [0, 1]$ s.t

$$d(f(x), f(y)) \leq cd(x, y) \quad \forall x, y \in X$$

Then f admits a unique fixed point.

Proof. Take $x_1 \in X$ and $x_{n+1} = f(x_n)$, we show that x_n is Cauchy. Indeed,

$$\begin{aligned}
 d(x_2, x_3) &= d(x_2, f(x_2)) \leq Cd(x_2, x_1)d(x_4, x_3) \leq Cd(x_3, x_2) \leq C^2 d(x_2, x_1): \\
 d(x_{n+1}, x_n) &\leq C^{n-1} d(x_2, x_1)
 \end{aligned}$$

Pick any $m > n$ then

$$\begin{aligned}
 d(x_m, x_n) &\leq d(x_{n+1}, x_n) + d(x_{n+2}, x_{n+1}) + \cdots + d(x_{m-1}, x_m) \\
 &\leq (C^{n-1} + C^n + \cdots + C^{m-2}) d(x_2, x_1) \\
 &= C^{n-1} (1 + C + C^2 + \cdots + C^{m-n-1}) d(x_2, x_1) \\
 &\leq \frac{C^{n-1}}{1 - C} d(x_2, x_1)
 \end{aligned}$$

Let $\varepsilon > 0$ then

$$\frac{C^{n-1}}{1 - C} d(x_2, x_1) \rightarrow 0$$

i.e $\exists N$ s.t $\forall n \geq N$ we have

$$\frac{C^{n-1}}{1 - C} d(x_2, x_1) < \varepsilon$$

Therefore for $m, n \geq N$ we have

$$d(x_m, x_n) < \varepsilon$$

Hence x_n is Cauchy therefore it converges to some x_0 (since X is complete). Moreover, L is Lipschitz continuous, and $x_{n+1} = f(x_n)$ let $n \rightarrow \infty$ we get

$$x_0 = f(x_0)$$

we have found a fixed point.

Let x_0, x'_0 be fixed points,

$$\begin{aligned} d(f(x_0), f(x'_0)) &\leq Cd(x_0, x'_0) \\ d(x_0, x'_0) &< d(x_0, x'_0) \end{aligned}$$

Therefore $d(x_0, x'_0) = 0$ hence $x'_0 = x_0$. □

Theorem 2.44 (Inverse function theorem). $U \subset \mathbb{R}^n$ open and $f: U \rightarrow \mathbb{R}^n$ a C^1 function and $x_0 \in U$ if $\det[Df(x_0)] \neq 0$ then $\exists O_1$ open neighborhood of x_0 and O_2 open neighborhood of $f(x_0)$ s.t $f: O_1 \rightarrow O_2$ is bijective. Moreover, $f^{-1} \in C_1(O_2)$ and

$$(Df^{-1})(f(x)) = (Df(x))^{-1} \quad \forall x \in O_1$$

Remark 2.45. It is enough to assume that $Df(x_0) = I_n$; suppose that we have the setting of the theorem

$$\det(Df(x_0)) \neq 0$$

Then take

$$g(x) = (Df(x_0))^{-1}f(x) = (Df(x_0)) \circ f(x)$$

Therefore by chain rule and the fact that $DT = T$, we have

$$Dg(x_0) = (Df(x_0))^{-1}(Df(x_0)) = I_n$$

Apply the IFT at g , we find $O_1 \ni x_0$ and $O_2 \ni g(x_0)$ s.t $g: O_1 \rightarrow O_2$ is a Diffeomorphism, but

$$g(x) = Df(x_0)^{-1} \circ f(x) \implies f(x) = Df(x_0)(g(x)): O_1 \rightarrow Df(x_0)O_2$$

Note that $Df(x_0)^{-1}$ is a linear transformation, hence continuous and contains $f(x_0)$. Moreover the composition of diffeomorphisms is a diffeomorphism.

More generally, suppose $h, h^{-1} \in C_1$ therefore

$$h^{-1} \circ h = \text{id}$$

therefore by chain rule

$$Dh^{-1}(h(x)) \circ Dh(x) = I_n \implies Dh^{-1}(h(x)) = (Dh(x))^{-1}$$