MA3101

Analysis III

Autumn 2021

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1 Euclidean spaces

1.1 \mathbb{R}^n as a vector space

We are familiar with the vector space \mathbb{R}^n , with the standard inner product

$$\langle \boldsymbol{x}, \boldsymbol{y} \rangle = x_1 y_1 + \dots + x_n y_n.$$

The standard norm is defined as

$$\|\boldsymbol{x} - \boldsymbol{y}\|^2 = \langle \boldsymbol{x} - \boldsymbol{y}, \boldsymbol{x} - \boldsymbol{y} \rangle = \sum_{k=1}^n (x_i - y_i)^2.$$

Exercise 1.1. What are all possible inner products on \mathbb{R}^n ?

Solution. Note that an inner product is a bilinear, symmetric map such that $\langle \boldsymbol{x}, \boldsymbol{x} \rangle \geq 0$, and $\langle \boldsymbol{x}, \boldsymbol{x} \rangle = 0$ if and only if $\boldsymbol{x} = \boldsymbol{0}$. Thus, an product map on \mathbb{R}^n is completely and uniquely determined by the values $\langle \boldsymbol{e}_i, \boldsymbol{e}_j \rangle = a_{ij}$. Let A be the $n \times n$ matrix with entries a_{ij} . Note that A is a real symmetric matrix with positive entries. Now,

$$\langle \boldsymbol{x}, \boldsymbol{e}_j \rangle = x_1 a_{1j} + \dots + x_n a_{nj} = \boldsymbol{x}^{\top} \boldsymbol{a}_j,$$

where a_j is the j^{th} column of A. Thus,

$$\langle \boldsymbol{x}, \boldsymbol{y} \rangle = \boldsymbol{x}^{\top} \boldsymbol{a}_1 y_1 + \dots + \boldsymbol{x}^{\top} \boldsymbol{a}_n y_n = \boldsymbol{x}^{\top} A \boldsymbol{y}.$$

Furthermore, any choice of real symmetric A with positive entries produces an inner product.

Theorem 1.1 (Cauchy-Schwarz). Given two vectors $v, w \in \mathbb{R}^n$, we have

$$|\langle \boldsymbol{v}, \boldsymbol{w} \rangle| \leq ||\boldsymbol{v}|| ||\boldsymbol{w}||.$$

Proof. This is trivial when w = 0. When $w \neq 0$, set $\lambda = \langle v, w \rangle / ||w||^2$. Thus,

$$0 \le \|\boldsymbol{v} - \lambda \boldsymbol{w}\|^2 = \|\boldsymbol{v}\|^2 - 2\lambda \langle \boldsymbol{v}, \boldsymbol{w} \rangle + \lambda^2 \|\boldsymbol{w}\|^2.$$

Simplifying,

$$0 \leq \|\boldsymbol{v}\|^2 - \frac{|\langle \boldsymbol{v}, \boldsymbol{w} \rangle|^2}{\|\boldsymbol{w}\|^2}.$$

This gives the desired result. Clearly, equality holds if and only if $v = \lambda w$.

Theorem 1.2 (Triangle inequality). Given two vectors $v, w \in \mathbb{R}^n$, we have

$$\|v + w\| \le \|v\| + \|w\|.$$

Proof. Write

$$\|v + w\|^2 = \|v\|^2 + 2\langle v, w \rangle + \|w\|^2 \le \|v\|^2 + 2|\langle v, w \rangle| + \|w\|^2.$$

Applying Cauchy-Schwarz gives

$$\|\boldsymbol{v} + \boldsymbol{w}\|^2 \le (\|\boldsymbol{v}\| + \|\boldsymbol{w}\|)^2.$$

Equality holds if and only if $\mathbf{v} = \lambda \mathbf{w}$ for $\lambda \geq 0$.

1.2 \mathbb{R}^n as a metric space

Our previous observations allow us to define the standard metric on \mathbb{R}^n , seen as a point set.

$$d(\boldsymbol{x}, \boldsymbol{y}) = \|\boldsymbol{x} - \boldsymbol{y}\|.$$

Definition 1.1. For any $\delta > 0$, the set

$$B_{\delta}(\boldsymbol{x}) = \{ \boldsymbol{y} \in \mathbb{R}^n : d(\boldsymbol{x}, \boldsymbol{y}) < \delta \}$$

is called the open ball centred at $x \in \mathbb{R}^n$ with radius δ . This is also called the δ neighbourhood of x.

Definition 1.2. A set U is open in \mathbb{R}^n if for every $\boldsymbol{x} \in U$, there exists an open ball $B_{\delta}(\boldsymbol{x}) \subset U$.

Remark. Every open ball in \mathbb{R}^n is open.

Remark. Both \emptyset and \mathbb{R}^n are open.

Definition 1.3. A set F is closed in \mathbb{R}^n if its complement $\mathbb{R}^n \setminus F$ is open in \mathbb{R}^n .

Remark. Both \emptyset and \mathbb{R}^n are closed.

Remark. Finite sets in \mathbb{R}^n are closed.

Theorem 1.3. Unions and finite intersections of open sets are open.

Corollary 1.3.1. Intersections and finite unions of closed sets are closed.

Definition 1.4. An interior point x of a set $S \subseteq \mathbb{R}^n$ is such that there is a neighbourhood of x contained within S.

Example. Every point in an open set is an interior point by definition. The interior of a set is the largest open set contained within it.

Definition 1.5. An exterior point x of a set $S \subseteq \mathbb{R}^n$ is an interior point of the complement $\mathbb{R}^n \setminus S$.

Definition 1.6. A boundary point of a set is neither an interior point, nor an exterior point.

Example. The boundary of the unit open ball $B_1(0) \subset \mathbb{R}^n$ is the sphere S^{n-1} .

Definition 1.7. A limit point x of a set $S \subseteq \mathbb{R}^n$ is such that every neighbourhood of x contains a point from S other than itself.

Definition 1.8. The closure of a set $S \subseteq \mathbb{R}^n$ is the union of S and its limit points.

Remark. The closure of a set is the smallest closed set containing it.

Lemma 1.4. Every open set in \mathbb{R}^n is a union of open balls.

Proof. Let $U \subseteq \mathbb{R}^n$ be open. Thus, for every $\boldsymbol{x} \in \mathbb{R}^n$, we can choose $\delta_x > 0$ such that $B_{\delta_x}(\boldsymbol{x}) \subset U$. The union of all such open balls is precisely the set U.

1.3 \mathbb{R}^n as a topological space

Definition 1.9. A topology on a set X is a collection τ of subsets of X such that

- 1. $\emptyset \in \tau$
- $2. X \in \tau$
- 3. Arbitrary union of sets from τ belong to τ .
- 4. Finite intersections of sets from τ belong to τ .

Sets from τ are called open sets.

Example. The Euclidean metric induces the standard topology on \mathbb{R}^n .

Example. The discrete topology on a set X is one where every singleton set is open. This is the topology induced by the discrete metric,

$$d_{\text{discrete}} \colon X \times X \to \mathbb{R}, \qquad (x,y) \mapsto \begin{cases} 0, & \text{if } x = y, \\ 1, & \text{if } x \neq y. \end{cases}$$

Example. Let X be an infinite set. The collection of sets consisting of \emptyset along with all sets A such that $X \setminus A$ is finite is a topology on X. This is called the Zariski topology.

Example. Consider the set of real numbers, and let τ be the collection \emptyset , \mathbb{R} , and all intervals (-x, +x) for x > 0. This constitutes a topology on \mathbb{R} , very different from the usual one.

This topology cannot be induced by a metric; it is not metrizable.

Consider the constant sequence of zeros. In this topology (\mathbb{R}, τ) , this sequence converges to *every* point in \mathbb{R} . Given any $\ell \in \mathbb{R}$, the open neighbourhoods of ℓ are precisely the sets \mathbb{R} and the open intervals (-x, +x) for $x > |\ell|$. The tail of the constant sequence of zeros is contained within every such neighbourhood of ℓ , hence $0 \to \ell$. Indeed, the element zero belongs to every open set apart from \emptyset in this topology.

Definition 1.10. A topological space is called Hausdorff if for every distinct $x, y \in X$, there exist disjoint neighbourhoods of x and y.

Example. Every metric space is Hausdorff. Given distinct x, y in a metric space (X, d), set $\delta = d(x, y)/3$ and consider the open balls $B_{\delta}(x)$ and $B_{\delta}(y)$.

Lemma 1.5. Every convergent sequence in a Hausdorff space has exactly one limit.

Proof. Consider a sequence $\{x_n\}_{n\in\mathbb{N}}$, and suppose that it converges to distinct x_1 and x_2 . Construct disjoint neighbourhoods U_1 and U_2 around x_1 and x_2 . Now, convergence implies that both U_1 and U_2 contain the tail of $\{x_n\}$, which is impossible since they are disjoint and hence contain no elements in common.

Definition 1.11. Given a topological space (X, τ) and a subset $Y \subseteq X$, the collection of sets $U \cap Y$ where $U \in \tau$ is a topology τ_Y on Y. We call this collection the subspace topology on Y, induced by the topology on X.

1.4 Compact sets in \mathbb{R}^n

Definition 1.12. A set $K \subset X$ in a topological space is compact if every open cover of K has a finite sub-cover. That is, for every collection if $\{U_{\alpha}\}_{{\alpha}\in A}$ of open sets such that K is contained in their union, there exists a finite sub-collection $U_{\alpha_1}, \ldots, U_{\alpha_k}$ such that K is also contained in their union.

Example. All finite sets are compact.

Example. Given a convergent sequence of real numbers $x_n \to x$, the collection $\{x_n\}_{n\in\mathbb{N}}\cup\{x\}$ is compact.

Example. In \mathbb{R}^n , compact sets are precisely those sets which are closed and bounded. This is the Heine-Borel Theorem.

Theorem 1.6. The closed intervals $[a,b] \subset \mathbb{R}$ are compact.

Remark. This can be extended to show that any k-cell $[a_1,b_1] \times \cdots \times [a_n,b_n] \subset \mathbb{R}^n$ is compact.

Proof. Let $\{U_{\alpha}\}_{{\alpha}\in A}$ be an open cover of [a,b], and suppose that $I_1=[a,b]$ has no finite subcover. Then, at least one of the intervals [a,(a+b)/2] and [(a+b)/2,b] must not have a finite sub-cover; pick one and call it I_2 . Similarly, one of the halves of I_2 must not have a finite

sub-cover; call it I_3 . In this process, we generate a sequence of closed intervals $I_1 \supset I_2 \supset \dots$, none of which have a finite sub-cover. The length of each interval is given by

$$|I_n| = 2^{-n+1} ||b - a|| \to 0.$$

Now, pick a sequence of points $\{x_n\}$ where each $x_n \in I_n$. Then, $\{x_n\}$ is a Cauchy sequence. To see this, given any $\epsilon > 0$, we can find sufficiently large n_0 such that $2^{-n_0+1}||b-a|| < \epsilon$. Thus, $x_n \in I_n \subset I_{n_0}$ for all $n \geq n_0$, which means that for any $m, n \geq n_0$, we have $x_m, x_n \in I_{n_0}$ forcing¹

$$||x_m - x_n|| \le |I_{n_0}| = 2^{-n_0 + 1} ||b - a|| < \epsilon.$$

From the completeness of \mathbb{R} , this sequence must converge in \mathbb{R} , specifically in [a,b]. Thus, $x_n \to x$ for some $x \in [a,b]$. It can also be seen that the limit $x \in I_n$ for all $n \in \mathbb{N}$; if not, say $x \notin I_{n_0}$, then $x \in [a,b] \setminus I_{n_0}$ which is open, hence there is an open interval such that $(x-\delta,x+\delta) \cap I_{n_0} = \emptyset$. However, I_{n_0} contains all $x_{n\geq n_0}$, thus this δ -neighbourhood of x would miss out a tail of $\{x_n\}$.

Now, pick the open set $U \in \{U_{\alpha}\}$ which covers the point x. Thus, $x \in U$ so U contains some non-empty open interval $(x - \delta, x + \delta)$ around x. Choose n_0 such that $2^{-n_0+1}||b-a|| < \delta$; this immediately gives $I_{n_0} \subseteq (x - \delta, x + \delta) \subset U$. This contradicts that fact that I_{n_0} has no finite sub-cover from $\{U_{\alpha}\}$, completing the proof.

Remark. The fact that Cauchy sequences in \mathbb{R}^n converge isn't immediately obvious; it is a consequence of the completeness of \mathbb{R}^n . Start by noting that \mathbb{R} has the Least Upper Bound property, from which the Monotone Convergence Theorem follows; every monotonic, bounded sequence of reals converges. It can also be shown that any sequence of reals with contain a monotone subsequence, from which it follows that every bounded sequence contains a convergent subsequence (Bolzano-Weierstrass). Finally, it can be shown that if a subsequence of a Cauchy sequence converges, then the entire sequence also converges to the same limit, giving us the desired result for \mathbb{R} . For sequence in \mathbb{R}^n , we may apply this coordinate-wise to obtain the result.

Lemma 1.7. Compact sets in \mathbb{R}^n are closed and bounded.

Proof. Consider a compact set $K \subset \mathbb{R}^n$. Let $x \in \mathbb{R}^n \setminus K$, and let $y \in K$. Since $x \neq y$, we choose open balls U_y around y and V_y around x such that $U_y \cap V_y = \emptyset$. Repeating this for all $y \in K$, we generate an open cover $\{U_y\}$ of K consisting of open balls. The compactness of K guarantees that this has a finite sub-cover, i.e. there is a finite set Y such that the collection $\{U_y\}_{y \in Y}$ covers X. As a result, the finite intersection of all V_y for $y \in Y$ is contained within $\mathbb{R}^n \setminus K$. Thus, x is in the exterior of K. Since x was chosen arbitrarily from $\mathbb{R}^n \setminus K$, we see that K is closed.

Now, consider the open cover $\{B_1(x)\}_{x\in K}$, and extract a finite sub-cover of unit open balls. The distance between any two points in K is at most the maximum distance between the centres of any two balls in our sub-cover, plus two.

Lemma 1.8. The intersection of a closed set and a compact set is compact.

$$|x_2 - x_1| = x_2 - x_1 \le b - a.$$

¹If $x_1, x_2 \in [a, b]$ with $x_1 < x_2$, note that $a < x_1 < x_2 < b$, so

Proof. Let $F \subseteq \mathbb{R}^n$ be closed and let $K \subseteq \mathbb{R}^n$ be compact. Suppose that the open cover $\{U_\alpha\}$ of $F \cap K$ has no finite sub-cover. Now the complement $U = F^c$ is open in \mathbb{R}^n , hence the collection $\{U_\alpha\} \cup \{U\}$ is an open cover of K, and hence must admit a finite sub-cover of K. In particular, this must be a finite sub-cover of $F \cap K$. However, we can remove the set U from this sub-cover since it shares no element with $F \cap K$; as a result, our sub-cover must be a finite sub-collection of sets U_α , contradicting our assumption. This shows that $F \cap K$ is compact.

Lemma 1.9 (Finite intersection property). Let $\{K_{\alpha}\}$ be a collection of compact sets in \mathbb{R}^n which have the property that any finite intersection of them is non-empty. Then,

$$\bigcap_{\alpha} K_{\alpha} \neq \emptyset.$$

Proof. Suppose to the contrary that the intersection of all K_{α} is empty. Fix an index β , and note that no element of K_{β} lies in every K_{α} . Set $J_{\alpha} = K_{\alpha}^{c}$, whence the collection $\{J_{\alpha} : \alpha \neq \beta\}$ is an open cover of K_{β} . This must admit a finite sub-cover $\{J_{\alpha_{1}}, \ldots, J_{\alpha_{k}}\}$ of K_{β} . Thus, we must have

$$K_{\beta}^c \cup J_{\alpha_1} \cup \cdots \cup J_{\alpha_k} = \mathbb{R}^n.$$

This immediately gives the contradiction

$$K_{\beta} \cap K_{\alpha_1} \cap \dots \cap K_{\alpha_k} = \emptyset.$$

Theorem 1.10 (Heine-Borel). Compact sets in \mathbb{R}^n are precisely those that are closed and bounded.

Proof. Given a compact set in \mathbb{R}^n , we have already shown that it must be closed and bounded. Next, if $F \subset \mathbb{R}^n$ is closed and bounded, it can be enclosed within a k-cell which we know is compact. Thus, F is the intersection of the closed set F and the compact k-cell, proving that F must be compact.

1.5 Continuous maps

Definition 1.13. A map $f: X \to Y$ is continuous if the pre-image of every open set from Y is open in X.

Lemma 1.11. A map $f: X \to Y$ is continuous if the pre-image of every closed set from Y is closed in X.

Theorem 1.12. The projection maps $\pi_i : \mathbb{R}^n \to \mathbb{R}$, $\boldsymbol{x} \mapsto x_i$ are continuous.

Proof. Let $U \subseteq \mathbb{R}$ be open; we claim that $\pi_i^{-1}(U)$ is open. Pick $\mathbf{x} \in \pi_i^{-1}(U)$, and note that $\pi_i(\mathbf{x}) = x_i \in U$. Thus, there exists $\delta > 0$ such that $(x_i - \delta, x_i + \delta) \subset U$. Now examine $B_{\delta}(\mathbf{x})$; for any point \mathbf{y} within this open ball, we have $d(\mathbf{x}, \mathbf{y}) < \delta$ hence

$$|x_i - y_i|^2 \le \sum_{k=1}^n (x_k - y_k)^2 = d(\boldsymbol{x}, \boldsymbol{y})^2 < \delta^2.$$

In other words, $\pi_i(\mathbf{y}) = y_i \in (x_i - \delta, x_i + \delta)$, hence $\pi_i B_{\delta}(\mathbf{x}) \subseteq (x_i - \delta, x_i + \delta) \subset U$. Thus, given arbitrary $\mathbf{x} \in \pi_i^{-1}(U)$, we have found an open ball $B_{\delta}(\mathbf{x}) \subset \pi_i^{-1}(U)$.

Lemma 1.13. Finite sums, products, and compositions of continuous functions are continuous.

Corollary 1.13.1. A function $f:[a,b] \to \mathbb{R}^n$ is continuous if and only if the components, $\pi_i \circ f$, are continuous.

Theorem 1.14. All polynomial functions of the coordinates in \mathbb{R}^n are continuous.

Example. The unit sphere $S^{n-1} \subset \mathbb{R}^n$ is closed. It is by definition the pre-image of the singleton closed set $\{1\}$ under the continuous map

$$\boldsymbol{x} \mapsto x_1^2 + \dots + x_n^2$$
.

Theorem 1.15. The continuous image of a compact set is compact.

Proof. Let $f: X \to Y$ be continuous, where Y is the image of the compact set X, and let $\{U_{\alpha}\}$ be an open cover of Y. Then, the collection $\{f^{-1}(U_{\alpha})\}$ is an open cover of X. Using the compactness of X, extract a finite sub-cover $f^{-1}(U_{\alpha_1}), \ldots, f^{-1}(U_{\alpha_k})$ of X. It follows that the collection $U_{\alpha_1}, \ldots, U_{\alpha_k}$ is a finite sub-cover of Y.

1.6 Connectedness

Definition 1.14. Let X be a topological space. A separation of X is a pair U, V of non-empty disjoint open subsets such that $X = U \cup V$.

Definition 1.15. A connected topological space is one which cannot be separated.

Lemma 1.16. A topological space X is connected if and only if the only sets which are both open and closed are \emptyset and X.

Example. The intervals $(a,b) \subset \mathbb{R}$ are connected. To see this, suppose that U,V is a separation of (a,b). Pick $x \in U$, $y \in V$, and without loss of generality let x < y. Define $S = [x,y] \cap U$, and set $c = \sup S$. It can be argued that $c \in (a,b)$, but $c \notin U$, $c \notin V$, using the properties of the supremum.

Theorem 1.17. The continuous image of a connected set is connected.

Proof. Let f be a continuous map on the connected set X, and let Y be the image of X. If U, V is a separation of Y, then it can be shown that $f^{-1}(U)$, $f^{-1}(V)$ constitutes a separation of X, which is a contradiction.

Definition 1.16. A path γ joining two points $x, y \in X$ is a continuous map $\gamma \colon [a, b] \to X$ such that $\gamma(a) = x$, $\gamma(b) = y$.

Definition 1.17. A set in X is path connected if given any two distinct points in X, there exists a path joining them.

Lemma 1.18. Every path connected set is connected.

Proof. Let X be path connected, and suppose that U, V is a separation of X. Then, pick $x \in U$, $y \in V$, and choose a path $\gamma \colon [0,1] \to X$ between x and y. The sets $f^{-1}(U)$ and $f^{-1}(V)$ separate the interval [0,1], which is a contradiction.

Example. All connected sets are not path connected. Consider the topologist's sine curve,

$$\left\{ \left(x, \sin \frac{1}{x} \right) : 0 < x \le 1 \right\} \cup \{ (0, 0) \}.$$

Definition 1.18. The ϵ neighbourhood of a set K in a metric space X is defined as

$$\bigcup_{a \in K} B_{\epsilon}(a) = \bigcup_{a \in K} \{x \in X : d(x, a) < \epsilon\}.$$

Exercise 1.2. Let $K \subseteq \mathbb{R}^n$ be compact, and define $f: \mathbb{R}^n \to \mathbb{R}$,

$$f(x) = \operatorname{dist}(x, K) = \inf_{a \in K} d(x, a).$$

Show that f is continuous on \mathbb{R}^n , and $f^{-1}(\{0\}) = K$.

Exercise 1.3. If $K \subseteq \mathbb{R}^n$ is compact and $K \cap L = \emptyset$, then

$$\operatorname{dist}(K,L) = \inf_{a \in K} \operatorname{dist}(a,L) > 0.$$

Exercise 1.4. If $K \subseteq \mathbb{R}^n$ is compact and U is an open set containing K, then there exists $\epsilon > 0$ such that U contains the ϵ neighbourhood of K.

Is the compactness of K necessary?

2 Differential calculus

2.1 Differentiability

Definition 2.1. Let $f:(a,b) \to \mathbb{R}^n$, and let $f_i = \pi_i \circ f$ be its components. Then, f is differentiable at $t_0 \in (a,b)$ if the following limit exists.

$$f'(t_0) = \lim_{h \to 0} \frac{f(t_0 + h) - f(t_0)}{h}.$$

Remark. The vector $f'(t_0)$ represents the tangent to the curve f at the point $f(t_0)$. The full tangent line is the parametric curve $f(t) + f'(t_0)(t - t_0)$.

Definition 2.2. Let $U \subseteq \mathbb{R}^n$ be open, and let $f: U \to \mathbb{R}^m$. Then, f is differentiable at $x \in U$ if there exists a linear transformation $\lambda \colon \mathbb{R}^n \to \mathbb{R}^m$ such that

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

The derivative of f at x is denoted by $\lambda = Df(x)$.

Remark. In a neighbourhood of x, we may approximate

$$f(x+h) \approx f(x) + Df(x)(h)$$
.

Remark. The statement that this quantity goes to zero means that each of the m components must also go to zero. For each of these limits, there are n axes along which we can let $h \to 0$. As a result, we obtain $m \times n$ limits, which allow us to identify the $m \times n$ components of the matrix representing the linear transformation λ (in the standard basis). These are the partial derivatives of f, and the matrix of λ is the Jacobian matrix of f evaluated at x.

Example. Let $T: \mathbb{R}^n \to \mathbb{R}^m$ be a linear map. By choosing $\lambda = T$, we see that T is differentiable everywhere, with DT(x) = T for every choice of $x \in \mathbb{R}^n$. This is made obvious by the fact that the best linear approximation of a linear map at some point is the map itself; indeed, the 'approximation' is exact.

Lemma 2.1. If $f: \mathbb{R}^n \to \mathbb{R}^m$ is differentiable at $x \in \mathbb{R}^n$, with derivative Df(x), then

- 1. f is continuous at x.
- 2. The linear transformation Df(x) is unique.

Proof. We prove the second part. Suppose that λ , μ satisfy the requirements for Df(x); it can be shown that $\lim_{h\to 0} (\lambda - \mu)h/\|h\| = 0$. Now, if $\lambda v \neq \mu v$ for some non-zero vector $v \in \mathbb{R}^n$, then

$$\lambda v - \mu v = \frac{\lambda(tv) - \mu(tv)}{\|tv\|} \cdot \|v\| \to 0,$$

a contradiction. \Box

2.2 Chain rule

Exercise 2.1. Let $T: \mathbb{R}^n \to \mathbb{R}^m$ be a linear transformation. Then, there exists M > 0 such that for all $x \in \mathbb{R}^n$, we have

$$||T\boldsymbol{x}|| \le M||\boldsymbol{x}||.$$

Solution. Set $v_i = T(e_i)$ where e_i are the standard unit basis vectors of \mathbb{R}^n . Then,

$$||Tx|| = ||\sum_{i} x_i v_i|| \le \sum_{i} ||x_i v_i|| \le \max ||v_i|| \sum_{i} |x_i|.$$

Since each $|x_i| \leq ||x||$, set $M = n \max ||v_i||$ and write

$$||Tx|| \le \max ||v_i|| \sum_i |x_i| \le \max ||v_i|| \cdot n||x|| = M||x||.$$

Theorem 2.2. Let $f: \mathbb{R}^n \to \mathbb{R}^m$, $g: \mathbb{R}^m \to \mathbb{R}^k$ where f is differentiable at $a \in \mathbb{R}^n$ and g is differentiable at $f(a) \in \mathbb{R}^m$. Then, $g \circ f$ is differentiable, with $D(g \circ f)(a) = Dg(f(a)) \circ Df(a)$. Note that this means that the Jacobian matrices simply multiply.

Proof. Set $b = f(a) \in \mathbb{R}^m$, $\lambda = Df(a)$, $\mu = Dg(f(a))$. Define

$$\varphi \colon \mathbb{R}^n \to \mathbb{R}^m, \qquad \varphi(x) = f(x) - f(a) - \lambda(x - a),$$

 $\psi \colon \mathbb{R}^m \to \mathbb{R}^k, \qquad \psi(y) = g(y) - g(b) - \mu(y - b).$

We claim that

$$\lim_{x \to a} \frac{g \circ f(x) - g \circ f(a) - \mu \circ \lambda(x - a)}{\|x - a\|} = 0.$$

Write the numerator as

$$g \circ f(x) - g \circ f(a) - \mu \circ \lambda(x - a) = \psi(f(x)) + \mu(\varphi(x)).$$

Note that

$$\lim_{x \to a} \frac{\varphi(x)}{\|x - a\|} = 0, \qquad \lim_{y \to b} \frac{\psi(y)}{\|y - b\|} = 0.$$

Thus, find M>0 such that $\|\mu(\varphi(x))\|\leq \|\varphi(x)\|$ for all $x\in\mathbb{R}^n$, hence

$$\lim_{x \to a} \frac{\|\mu(\varphi(x))\|}{\|x - a\|} \le \lim_{x \to a} \frac{M\|\varphi(x)\|}{\|x - a\|} = 0.$$

Now write

$$\lim_{f(x)\to b}\frac{\psi(f(x))}{\|f(x)-b\|}=0,$$

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hence for any $\epsilon > 0$, there is a neighbourhood of b on which

$$\|\psi(f(x))\| \le \epsilon \|f(x) - b\| = \epsilon \|\varphi(x) + \lambda(x - a)\|.$$

Apply the triangle inequality and find M' > 0 such that

$$\|\psi(f(x))\| \le \epsilon \|\varphi(x)\| + \epsilon M' \|x - a\|.$$

Thus,

$$\lim_{x \to a} \frac{\|\psi(f(x))\|}{\|x - a\|} \le \lim_{x \to a} \frac{\epsilon \|\varphi(x)\|}{\|x - a\|} + \epsilon M' = \epsilon M'.$$

Since $\epsilon > 0$ was arbitrary, this limit is zero, completing the proof.

2.3 Partial derivatives

Definition 2.3. Let $U \subseteq \mathbb{R}^n$ be open, and let $f: U \to \mathbb{R}$. The partial derivative of f with respect to the coordinate x_i at some $a \in U$ is defined by the following limit, if it exists.

$$\frac{\partial f}{\partial x_j}(a) = \lim_{h \to 0} \frac{f(a + h\mathbf{e}_j) - f(a)}{h}.$$

Lemma 2.3. If $f: U \to \mathbb{R}$ is differentiable at a point $a \in \mathbb{R}^n$, then

$$Df(a)(x_1,...,x_n) = x_1 \frac{\partial f}{\partial x_1}(a) + \cdots + x_n \frac{\partial f}{\partial x_n}(a).$$

Example. Consider

$$f: \mathbb{R}^2 \to \mathbb{R}, \qquad (x,y) \mapsto \begin{cases} xy/(x^2 + y^2), & \text{if } (x,y) \neq (0,0), \\ 0, & \text{if } (x,y) = (0,0). \end{cases}$$

Note that f is not differentiable at (0,0); it is not even continuous there. However, both partial derivatives of f exist at (0,0).

Lemma 2.4. If $f: \mathbb{R}^n \to \mathbb{R}^m$ is differentiable at $a \in \mathbb{R}^n$, then the matrix representation of Df(a) in the standard basis is given by

$$[Df(a)] = \left[\frac{\partial f_i}{\partial x_j}(a)\right]_{ij}.$$

Lemma 2.5. Let $f: \mathbb{R}^n \to \mathbb{R}^m$ be differentiable at $a \in \mathbb{R}^n$, and let $g: \mathbb{R}^m \to \mathbb{R}^k$ be differentiable at $f(a) \in \mathbb{R}^m$. Then, the matrix representation of $D(g \circ f)(a)$ in the standard basis is the product

$$[D(g \circ f)(a)] = [Dg(f(a))][Df(a)] = \left[\sum_{\ell=1}^{m} \frac{\partial g_i}{\partial y_\ell} \frac{\partial f_\ell}{\partial x_j}\right]_{ij}.$$

In other words,

$$\frac{\partial}{\partial x_j}(g \circ f)_i(a) = \sum_{\ell=1}^m \frac{\partial g_i}{\partial y_\ell}(f(a)) \frac{\partial f_\ell}{\partial x_j}(a).$$

Example. Let $f: \mathbb{R}^2 \to \mathbb{R}$ be differentiable, and let $\Gamma(f) = \{(x, y, f(x, y)) : x, y \in \mathbb{R}\}$ be the graph of f. Now, let $\gamma: [-1, 1] \to \Gamma(f)$ be a differentiable curve, represented by

$$\gamma(t) = (g(t), h(t), f(g(t), h(t))).$$

Then, we can compute the derivative

$$\gamma'(a) = \left(g'(a), h'(a), g'(a) \frac{\partial f}{\partial x} + h'(a) \frac{\partial f}{\partial y} \Big|_{(g(a), h(a))} \right)$$

Exercise 2.2. Consider the inner product map, $\langle \cdot, \cdot \rangle \colon \mathbb{R}^n \times \mathbb{R}^n \to \mathbb{R}$. What is its derivative? *Solution.* We treat the inner product as a map $g \colon \mathbb{R}^{2n} \to \mathbb{R}$, which acts as

$$\langle \boldsymbol{x}, \boldsymbol{y} \rangle \cong g(x_1, \dots, x_n, y_1, \dots, y_n) = x_1 y_1 + \dots + x_n y_n.$$

Now, note that

$$\frac{\partial g}{\partial x_i} = y_i, \qquad \frac{\partial g}{\partial y_i} = x_i.$$

Thus,

$$Dg(\boldsymbol{a}, \boldsymbol{b})(\boldsymbol{x}, \boldsymbol{y}) = \sum_{i=1}^{n} x_i \frac{\partial g}{\partial x_i}(\boldsymbol{a}, \boldsymbol{b}) + \sum_{i=1}^{n} y_i \frac{\partial g}{\partial y_i}(\boldsymbol{a}, \boldsymbol{b})$$
$$= \sum_{i=1}^{n} x_i b_i + \sum_{i=1}^{n} y_i a_i$$
$$= \langle \boldsymbol{x}, \boldsymbol{b} \rangle + \langle \boldsymbol{y}, \boldsymbol{a} \rangle.$$

In other words, the matrix representation of the derivative of the inner product map at the point (a, b) is given by $[b^{\top} a^{\top}]$.

Exercise 2.3. Let $\gamma : \mathbb{R} \to \mathbb{R}^n$ be a differentiable curve. What is the derivative of the real map $t \mapsto \|\gamma(t)\|^2$?

Solution. We write this map as $t \mapsto \langle \gamma(t), \gamma(t) \rangle$. Consider the scheme

$$\mathbb{R} \to \mathbb{R}^{2n} \to \mathbb{R}, \qquad t \mapsto \begin{bmatrix} \gamma(t) \\ \gamma(t) \end{bmatrix} \mapsto \langle \gamma(t), \gamma(t) \rangle.$$

Pick a point $t \in \mathbb{R}$, whence the derivative of the map at t is

$$\begin{bmatrix} \gamma(t)^\top & \gamma(t)^\top \end{bmatrix} \begin{bmatrix} \gamma'(t) \\ \gamma'(t) \end{bmatrix} = 2\langle \gamma(t), \gamma'(t) \rangle.$$

Remark. Consider the surface $S^{n-1} \subset \mathbb{R}^n$, and pick an arbitrary differentiable curve $\gamma \colon \mathbb{R} \to S^{n-1}$. Now, the tangent vector $\gamma'(t)$ is tangent to the sphere S^{n-1} at any point $\gamma(t)$. We claim that this tangent drawn at $\gamma(t)$ is always perpendicular to the position vector $\gamma(t)$. This is made trivial by our exercise: the map $t \mapsto ||\gamma(t)||^2 = 1$ is a constant map since γ is a curve on the unit sphere. This means that it has zero derivative, forcing $\langle \gamma(t), \gamma'(t) \rangle = 0$.

2.3.1 Directional derivatives

Definition 2.4. Let $U \subseteq \mathbb{R}^n$ be open, and let $f: U \to \mathbb{R}$. The directional derivative of f along a direction $v \in \mathbb{R}^n$ at a point $a \in U$ is defined by the following limit, if it exists.

$$\nabla_v f(a) = \lim_{h \to 0} \frac{f(a+h\boldsymbol{v}) - f(a)}{h}.$$

Example. Consider

$$f: \mathbb{R}^2 \to \mathbb{R}, \qquad (x,y) \mapsto \begin{cases} x^3/(x^2 + y^2), & \text{if } (x,y) \neq (0,0), \\ 0, & \text{if } (x,y) = (0,0). \end{cases}$$

Note that f is not differentiable at (0,0). However, all directional derivatives derivatives of f exist at (0,0). Indeed, consider a direction $(\cos \theta, \sin \theta)$, and examine the limit

$$\lim_{t \to 0} \frac{1}{t} \left[f(t\cos\theta, t\sin\theta) - f(0, 0) \right] = \cos^3\theta.$$

Definition 2.5. Let $f: \mathbb{R}^n \to \mathbb{R}$ be differentiable. The gradient of f is defined as the map

$$\nabla f \colon \mathbb{R}^n \to \mathbb{R}^n, \qquad x \mapsto \left[\frac{\partial f}{\partial x_i}(x) \right]_i.$$

Remark. The gradient at a point $x \in \mathbb{R}^n$ is thought of as a vector. In contrast, the derivative is thought of as a linear transformation. Otherwise, we see that $\nabla f(x) = [Df(x)]$.

Definition 2.6. Let $C^1(\mathbb{R}^n)$ be the set of real-valued differentiable functions on \mathbb{R}^n . Fix a point $a \in \mathbb{R}^n$, then fix a tangent vector $v \in \mathbb{R}^n$. Then, the map

$$\nabla_v \colon C^1(\mathbb{R}^n) \to \mathbb{R}, \qquad f \mapsto Df(a)(v)$$

is a linear functional. The quantity $\nabla_v f$ is called the directional derivative of f in the direction v at the point a.

Remark. We can represent ∇_v as the operator

$$\nabla_v(\cdot) = D(\cdot)(a)(v) = \sum_i v_i \frac{\partial}{\partial x_i} \Big|_a = v \cdot \nabla(\cdot).$$

Lemma 2.6. The directional derivatives ∇_v form a vector space called the tangent space, attached to the point $a \in \mathbb{R}^n$. This can be identified with the vector space \mathbb{R}^n by the natural map $\nabla_v \mapsto v$. The standard basis can be informally denoted by the vectors

$$\nabla_{\boldsymbol{e}_1} \equiv \frac{\partial}{\partial x_1}, \dots, \nabla_{\boldsymbol{e}_n} \equiv \frac{\partial}{\partial x_n}.$$

2.3.2 Differentiation on manifolds *

Definition 2.7. A homeomorphism is a continuous, bijective map whose inverse is also continuous.

Lemma 2.7. Let $f: \mathbb{R}^n \to \mathbb{R}$ be continuous. Denote the graph of f as

$$\Gamma(f) = \{(x, f(x)) : x \in \mathbb{R}^n\}.$$

Then, $\Gamma(f)$ is a smooth manifold.

Proof. Consider the homeomorphism

$$\varphi \colon \Gamma(f) \to \mathbb{R}^n, \qquad (x, f(x)) \mapsto x.$$

This is clearly bijective, continuous (restriction of a projection map), with a continuous inverse (from the continuity of f). Call this homeomorphism φ a coordinate map on $\Gamma(f)$.

Definition 2.8. Let $f: M \to \mathbb{R}$ where M is a smooth manifold, with a coordinate map $\varphi \colon M \to \mathbb{R}^n$. We say that f is differentiable at a point $a \in M$ if $f \circ \varphi^{-1} \colon \mathbb{R}^n \to \mathbb{R}$ is differentiable at $\varphi(a)$.

Definition 2.9. Let $f: M \to \mathbb{R}$ where M is a smooth manifold, let $\varphi: M \to \mathbb{R}^n$ be a coordinate map, and let $a \in M$. Let $\gamma: \mathbb{R} \to M$ be a curve such that $\gamma(0) = a$, and further let γ be differentiable in the sense that $\varphi \circ \gamma: \mathbb{R} \to \mathbb{R}^n$ is differentiable. The directional derivative of f at a along γ is defined as

$$\left. \frac{d}{dt} f(\gamma(t)) \right|_{t=0} = \lim_{h \to 0} \frac{f(\gamma(t+h)) - f(\gamma(t))}{h} \Big|_{t=0}.$$

Note that we are taking the derivative of $f \circ \gamma \colon \mathbb{R} \to \mathbb{R}$ in the conventional sense.

Lemma 2.8. Let γ_1 and γ_2 be two curves in M such that $\gamma_1(0) = \gamma_2(0) = a$, and

$$\frac{d}{dt}\varphi \circ \gamma_1(t)\Big|_{t=0} = \frac{d}{dt}\varphi \circ \gamma_2(t)\Big|_{t=0}.$$

In other words, γ_1 and γ_2 pass through the same point a at t=0, and have the same velocities there. Then, the directional derivatives of f at a along γ_1 and γ_2 are the same.

Definition 2.10. Let M be a smooth manifold, and let $a \in M$. Consider the following equivalence relation on the set of all curves γ in M such that $\gamma(0) = a$.

$$\gamma_1 \sim \gamma_2 \quad \Longleftrightarrow \quad \frac{d}{dt} \varphi \circ \gamma_1(t) \Big|_{t=0} = \frac{d}{dt} \varphi \circ \gamma_2(t) \Big|_{t=0}.$$

Each resultant equivalence class of curves is called a tangent vector at $a \in M$. Note that all these curves in a particular equivalence class pass through a with the same velocity vector.

The collection of all such tangent vectors, i.e. the space of all curves through a modulo the equivalence relation which identifies curves with the same velocity vector through a, is called the tangent space to M at a, denoted T_aM .

Remark. Each tangent vector $v \in T_aM$ acts on a differentiable function $f: M \to \mathbb{R}$ yielding a (well-defined) directional derivative at a.

$$v: C^1(M) \to \mathbb{R}, \qquad f \mapsto \frac{d}{dt} f(\gamma_v(t)) \Big|_{t=0}.$$

Thus, the tangent space represents all the directions in which taking a derivative of f makes sense.

Remark. The tangent space T_aM is a vector space. Upon fixing f, the map $Df(a): T_aM \to \mathbb{R}$, $v \mapsto vf(a)$ is a linear functional on the tangent space.

Remark. Given a tangent vector $v \in T_aM$, it can be identified with its corresponding velocity vector in \mathbb{R}^n . Thus, the tangent space T_aM can be identified with the geometric tangent plane drawn to the manifold M at the point a.

2.4 Mean value theorem

Consider a differentiable function $f: \mathbb{R}^n \to \mathbb{R}$, and fix $a \in \mathbb{R}^n$. Define the functions

$$g_i \colon \mathbb{R} \to \mathbb{R}, \qquad g_i(x) = f(a_1, \dots, a_{i-1}, x, a_{i+1}, \dots, a_n).$$

Then, each g_i is differentiable, with

$$g_i'(x) = \frac{\partial f}{\partial x_i}(a_1, \dots, a_{i-1}, x, a_{i+1}, \dots, a_n).$$

By applying the Mean Value Theorem on some interval [c,d], we can find $\alpha \in (c,d)$ such that $g_i(d) - g_i(c) = g_i'(\alpha)(d-c)$. In other words,

$$f(\ldots,d,\ldots)-f(\ldots,c,\ldots)=\frac{\partial f}{\partial x_i}(\ldots,\alpha,\ldots)(d-c).$$

Theorem 2.9. Let $f: \mathbb{R}^n \to \mathbb{R}^m$ and $a \in \mathbb{R}^n$. Then, f is differentiable at a if all the partial derivatives $\partial f/\partial x_j$ exist in a neighbourhood of a and are continuous at a.

Proof. Without loss of generality, let m=1. We claim that

$$\lim_{h \to 0} \frac{1}{\|h\|} \|f(a+h) - f(a) - \sum_{i=0}^{n} \frac{\partial f}{\partial x_i}(a) h_i\| = 0.$$

Examine

$$f(a+h) - f(a) = f(a_1 + h_1, \dots, a_n + h_n) - f(a_1, \dots, a_n)$$

$$= f(a_1 + h_1, \dots, a_n + h_n) - f(a_1 + h_1, \dots, a_{n-1} + h_{n-1}, a_n) +$$

$$f(a_1 + h_1, \dots, a_{n-1} + h_{n-1}, a_n) - f(a_1 + h_1, \dots, a_{n-1}, a_n) +$$

$$\vdots$$

$$f(a_1 + h_1, a_2, \dots, a_n) - f(a_1, \dots, a_n)$$

$$= \frac{\partial f}{\partial x_n}(c_n)h_n + \dots + \frac{\partial f}{\partial x_1}(c_1)h_1.$$

The last step follows from the Mean Value Theorem. As $h \to 0$, each $c_i \to a$. Thus,

$$\frac{1}{\|h\|} \|f(a+h) - f(a) - \sum_{i=0}^{n} \frac{\partial f}{\partial x_{i}}(a)h_{i}\| = \frac{1}{\|h\|} \|\sum_{i=0}^{n} \left(\frac{\partial f}{\partial x_{i}}(c_{i}) - \frac{\partial f}{\partial x_{i}}(a)\right) h_{i}\|$$

$$\leq \sum_{i=0}^{n} \left|\frac{\partial f}{\partial x_{i}}(c_{i}) - \frac{\partial f}{\partial x_{i}}(a)\right| \frac{|h_{i}|}{\|h\|}$$

$$\leq \sum_{i=0}^{n} \left|\frac{\partial f}{\partial x_{i}}(c_{i}) - \frac{\partial f}{\partial x_{i}}(a)\right|.$$

Taking the limit $h \to 0$, observe that $\partial f/\partial x_i(c_i) \to \partial f/\partial x_i(a)$ by the continuity of the partial derivatives, completing the proof.

Corollary 2.9.1. All polynomial functions on \mathbb{R}^n are differentiable.

Theorem 2.10. Let $f: \mathbb{R}^n \to \mathbb{R}$ be differentiable with continuous partial derivatives, and let $a \in \mathbb{R}^n$ be a point of local maximum. Then, Df(a) = 0.

Proof. We need only show that each

$$\frac{\partial f}{\partial x_i}(a) = 0.$$

This must be true, since a is also a local maximum of each of the restrictions g_i as defined earlier.

2.5 Inverse and implicit function theorems

Theorem 2.11 (Inverse function theorem). Let $f: \mathbb{R}^n \to \mathbb{R}^n$ be continuously differentiable on a neighbourhood of $a \in \mathbb{R}^n$, and let $\det(Df(a)) \neq 0$. Then, there exist neighbourhoods U of a and W of f(a) such that the restriction $f: U \to W$ is invertible. Furthermore, f^{-1} is continuous on U and differentiable on U.

Lemma 2.12. Consider a continuously differentiable function $f: \mathbb{R}^n \to \mathbb{R}$, and let M denote the surface defined by the zero set of f. Then, M can be represented as the graph of a differentiable function $h: \mathbb{R}^{n-1} \to \mathbb{R}$ at those points where $Df \neq 0$.

Proof. Without loss of generality, suppose that $\partial f/\partial x_n \neq 0$ at some point $a \in M$. It can be shown that the map

$$F: \mathbb{R}^n \to \mathbb{R}^n, \quad x \mapsto (x_1, x_2, \dots, x_{n-1}, f(x))$$

is invertible in a neighbourhood W of a, with a continuous and differentiable inverse of the form

$$G: \mathbb{R}^n \to \mathbb{R}^n, \quad u \mapsto (u_1, u_2, \dots, u_{n-1}, g(u)).$$

Since $F \circ G$ must be the identity map on W, we demand

$$(x_1, x_2, \dots, x_{n-1}, f(x_1, x_2, \dots, x_{n-1}, g(x))) = (x_1, x_2, \dots, x_{n-1}, x_n).$$

Thus, the zero set of f in this neighbourhood of a satisfies $x_n = 0$, hence

$$f(x_1, x_2, \dots, x_{n-1}, g(x_1, x_2, \dots, x_{n-1}, 0)) = 0.$$

In other words, the part of the surface M in the neighbourhood of a is precisely the set of points

$$(x_1, x_2, \ldots, x_{n-1}, g(x_1, x_2, \ldots, x_{n-1}, 0)).$$

Simply set

$$h: \mathbb{R}^{n-1} \to \mathbb{R}, \qquad x \mapsto g(x_1, x_2, \dots, x_{n-1}, 0),$$

whence the surface M is locally represented by the graph of h.

Remark. Note that by using

$$f(x_1,\ldots,x_{n-1},h(x_1,\ldots,x_{n-1}))=0$$

on the surface, we can use the chain rule to conclude that for all $1 \le i < n$, we have

$$\frac{\partial f}{\partial x_i}(a) + \frac{\partial f}{\partial x_n}(a) \frac{\partial h}{\partial x_i}(a_1, \dots, a_{n-1}) = 0.$$

Theorem 2.13 (Implicit function theorem). Let $f: \mathbb{R}^n \times \mathbb{R}^m \to \mathbb{R}^m$ be continuously differentiable in an open set containing (a,b), with f(a,b) = 0. Let $\det(\partial f^j/\partial x_{n+k}(a,b)) \neq 0$. Then, there exists an open set $U \subset \mathbb{R}^n$ containing a, an open set $V \subset \mathbb{R}^m$ containing b, and a differentiable function $g: U \to V$ such that f(x,g(x)) = 0.

Remark. The condition on the determinant can be rephrased as rank Df(a,b) = m.

Theorem 2.14. Let $f: \mathbb{R}^n \to \mathbb{R}$ be continuously differentiable, and let M be the surface defined by its zero set. Furthermore, let $\nabla f(a) \neq 0$ for some $a \in M$; thus, M can be locally represented by a graph on \mathbb{R}^{n-1} . Then, $\nabla f(a)$ is normal to the tangent vectors drawn at a to M; in fact, the perpendicular space of $\nabla f(a)$ is precisely the tangent space T_aM .

Proof. Consider a tangent vector drawn at a to M, represented by the differentiable curve $\gamma \colon \mathbb{R} \to M$, $\gamma(0) = a$; note that we use the identification $\gamma'(0) = v \in \mathbb{R}^n$. Then, calculate

$$\frac{d}{dt}f(\gamma(t))\Big|_{t=0} = Df(\gamma(0))(\gamma'(0)) = Df(a)(v).$$

On the other hand, we have $f(\gamma(t)) = 0$ identically. Thus,

$$v \cdot \nabla f(a) = Df(a)(v) = 0.$$

2.6 Taylor's theorem

Theorem 2.15. Let $f: \mathbb{R}^n \to \mathbb{R}^n$ have continuous second order partial derivatives. Then,

$$\frac{\partial^2 f}{\partial x_i \partial x_j} = \frac{\partial^2 f}{\partial x_j \partial x_i}.$$

Theorem 2.16. Let $f: \mathbb{R}^2 \to \mathbb{R}$ have continuous second order partial derivatives, and let $(x_0, y_0) \in \mathbb{R}^2$. Then, there exists $\epsilon > 0$ such that for all $||(x - x_0, y - y_0)|| < \epsilon$,

$$f(x,y) = f(x_0, y_0) + \frac{\partial f}{\partial x}(x - x_0) + \frac{\partial f}{\partial y}(y - y_0)$$
$$+ \frac{1}{2}\frac{\partial^2 f}{\partial x^2}(x - x_0)^2 + \frac{1}{2}\frac{\partial^2 f}{\partial y^2}(y - y_0)^2$$
$$+ \frac{\partial^2 f}{\partial x \partial y}(x - x_0)(y - x_0) + R(x, y),$$

where as $(x,y) \to (x_0,y_0) \to 0$, the remainder term vanishes as

$$\frac{|R(x,y)|}{\|(x-x_0,y-y_0)\|^2}\to 0.$$

All partial derivatives here are evaluated at (x_0, y_0) .

Proof. This follows from applying the Taylor's Theorem in one variable to the real function $g: \mathbb{R} \to \mathbb{R}, t \mapsto f((1-t)(x_0, y_0) + t(x, y)).$

2.7 Critical points and extrema

Definition 2.11. We say that $a \in \mathbb{R}^n$ is a critical point of $f: \mathbb{R}^n \to \mathbb{R}$ if all $\partial f/\partial x^j = 0$ there.

Lemma 2.17. All points of extrema of a differentiable function are critical points.

Proof. We already know that Df(a) = 0 where a is either a point of maximum or minimum.

Example. In order to find a point of extrema of a C^2 -smooth function $f: \mathbb{R}^2 \to \mathbb{R}$, we first identify a critical point (x_0, y_0) . Next, we must find a neighbourhood of (x_0, y_0) which contains no other critical points – to do this, apply Taylor's Theorem. Indeed, we see that

$$f(x,y) = f(x_0, y_0) + A(x - x_0)^2 + 2B(x - x_0)(y - y_0) + C(y - y_0)^2 + R_2.$$

For non-degeneracy of solutions, we demand $AC - B^2 \neq 0$, i.e. at (x_0, y_0) , we want

$$\left[\frac{\partial^2 f}{\partial x \partial y}\right]^2 \neq \frac{\partial f}{\partial x} \frac{\partial f}{\partial y}.$$

If $AC-B^2>0$ and $\partial^2 f/\partial x^2>0$, then we have found a point of minima; if $\partial^2 f/\partial x^2<0$, then we have found a point of maximum. If $AC-B^2<0$, then we have found a saddle point.

Example. Suppose that we wish to maximize the function $f: \mathbb{R}^2 \to \mathbb{R}$, given an equation of constraint g = 0, where $g: \mathbb{R}^2 \to \mathbb{R}$. Using the method of Lagrange multipliers, we look for solutions of the system

$$\begin{cases} \nabla f(x,y) + \lambda \nabla g(x,y) = 0, \\ g(x,y) = 0. \end{cases}$$

3 Integral calculus

3.1 Path integrals

Definition 3.1. A closed curve $\gamma \colon [a,b] \to \mathbb{R}^n$ is closed if $\gamma(a) = \gamma(b)$. It is called simple if it has no self intersections.

Definition 3.2. Let $p,q\colon U\to\mathbb{R}$ be continuous, where $U\subseteq\mathbb{R}^2$ is an open set, and let $\gamma\colon [a,b]\to U$ be piecewise smooth, i.e. smooth on (a,b) at all but finitely many points. Then, we define

$$\int_{\gamma} p \, dx + q \, dy = \int_a^b p(\gamma(t)) \, \gamma_1'(t) + q(\gamma(t)) \, \gamma_2'(t) \, dt.$$

Lemma 3.1. Let $\gamma: [a,b] \to \mathbb{R}^2$ be a smooth curve, and let $\varphi: [c,d] \to [a,b]$ be smooth, such that $\varphi(c) = a$ and $\varphi(d) = b$. Then, the composition $\gamma: \varphi: [c,d] \to \mathbb{R}^2$ is a smooth curve, and

$$\int_{\gamma \circ \varphi} p \, dx + q \, dy = \int_{c}^{d} \left[p(\gamma \circ \varphi(s)) \, \gamma_{1}'(\varphi(s)) + p(\gamma \circ \varphi(s)) \, \gamma_{2}'(\varphi(s)) \right] \varphi'(s) \, ds.$$

By substituting the parameter $\varphi(s) = t$, $\varphi'(s) ds = dt$, we retrieve

$$\int_{\gamma \circ \varphi} p \, dx + q \, dy = \int_a^b p(\gamma(t)) \, \gamma_1'(t) + q(\gamma(t)) \, \gamma_2'(t) \, dt = \int_{\gamma} p \, dx + q \, dy.$$

Theorem 3.2. Let $p, q: U \to \mathbb{R}^2$ be continuous, and let $\gamma: [a, b] \to U$ be a smooth curve. The integral

$$\int_{\gamma} p \, dx + q \, dy$$

depends only on the endpoints of γ if and only if there exists $u: U \to \mathbb{R}$ such that

$$p = \frac{\partial u}{\partial x}, \qquad q = \frac{\partial u}{\partial y}.$$

In other words, we demand that that the vector field (p,q) be the gradient of u.

Proof. First suppose that there exists u such that $\nabla u = (p,q)$. Then,

$$\int_{\gamma} p \, dx + q \, dy = \int_{a}^{b} \frac{\partial u}{\partial x} (\gamma(t)) \gamma_{1}'(t) + \frac{\partial u}{\partial y} (\gamma(t)) \gamma_{2}'(t) \, dt.$$

The chain rule shows that this is simply

$$\int_{a}^{b} \frac{d}{dt} u(\gamma(t)) dt = u(\gamma(b)) - u(\gamma(a)).$$

Conversely, suppose that the given integral depends only on the endpoints of γ . Given two points $\alpha, \beta \in U$, we construct a path from α to β by travelling only along the axes. Pick $(x,y) \in U$, and define $u: U \to \mathbb{R}$,

$$u(x,y) = \int_{\gamma} p \, dx + q \, dy,$$

where γ is such a polygonal path from a fixed point α to (x, y). Note that u is well-defined by the independence of choice of path γ .

Example. Let $f: \mathbb{R}^2 \to \mathbb{R}^2$ be continuous, and let γ be a smooth curve in \mathbb{R}^2 . We may denote

$$\int_{\gamma} \cdot ds = \int_{\gamma} f^1 \, dx + f^2 \, dy.$$

3.2 Multiple integrals

Definition 3.3. Let $f: [a_1, b_1] \times [a_2, b_2] \to \mathbb{R}$ be continuous. Now, let P be a partition of the rectangular domain into $n \times n$ sub-rectangles, and define

$$M_{ij} = \sup_{[x_i, x_{i+1}] \times [y_j, y_{j+1}]} f(x, y), \qquad y_{ij} = \inf_{[x_i, x_{i+1}] \times [y_j, y_{j+1}]} f(x, y).$$

We also define,

$$U(f,P) = \sum_{i,j} M_{ij}(x_i - x_{i-1})(y_j - y_{j-1}), \qquad L(f,P) = \sum_{i,j} m_{ij}(x_i - x_{i-1})(y_j - y_{j-1}).$$

Finally define the upper and lower sums

$$U(f) = \sup_{P} U(f, P), \qquad L(f) = \sup_{P} L(f, P).$$

Then, f is Riemann integrable if U(f) = L(f), and this integral is denoted by

$$\int_{[a_1,b_1]\times[a_2,b_2]} f$$

Remark. This definition naturally extends to integrals over any k-cell.

Definition 3.4. A measure zero set $E \subset \mathbb{R}^n$ is such that given any $\epsilon > 0$, there exists a countable collection of rectangles $\{A_j\}$ such that their union contains E, and the sum of their volumes is less than ϵ .

Example. Any countable subset of \mathbb{R}^n has measure zero.

Example. Any line in \mathbb{R}^2 , plane in \mathbb{R}^3 , etc. has measure zero.

Lemma 3.3. The countable union of measure zero sets has measure zero.

Theorem 3.4. A bounded function $f: A \to \mathbb{R}$ where $A \subset \mathbb{R}^n$ is a rectangle is integrable if and only if for every $\epsilon > 0$, there exists a partition P of A such that $U(f, P) - L(f, P) < \epsilon$.

Theorem 3.5. Let $f: A \to \mathbb{R}$ be bounded, where $A \subset \mathbb{R}^n$ is a rectangle. Then, f is Riemann integrable if and only if its set of discontinuities has measure zero.

Theorem 3.6. Let $f: A_1 \times A_2 \to \mathbb{R}$ be continuous, where $A_1 \subset \mathbb{R}^n$ and $A_2 \subset \mathbb{R}^m$ are closed rectangles. Then, we can write

$$\int_{A_1 \times A_2} f = \int_{A_2} \left(\int_{A_1} f(x, y) \, dx \right) \, dy = \int_{A_1} \left(\int_{A_2} f(x, y) \, dy \right) \, dx.$$

Theorem 3.7 (Green's theorem). Let γ be a smooth simple closed curve in \mathbb{R}^2 oriented counter-clockwise, and let Ω be the region enclosed by γ . If $p,q:\Omega \to \mathbb{R}$ have continuous partial derivatives, then

$$\int_{\gamma} p \, dx + q \, dy = \iint_{\Omega} \left(\frac{\partial q}{\partial x} - \frac{\partial p}{\partial y} \right) \, dx \, dy.$$

Example. Let $\gamma \colon [0, 2\pi] \to \mathbb{R}$, $\gamma(t) = (\cos t, \sin t)$. Then, γ enclosed the unit disc in \mathbb{R}^2 . To calculate its area, we can set p = 0, q = x, giving

$$\iint_{\Omega} dx \, dy = \int_{0}^{2\pi} \cos^2 t \, dt = \pi.$$

Another option is to set p = -y/2, q = x/2, giving

$$\iint_{\Omega} dx \, dy = \frac{1}{2} \int_{0}^{2\pi} \cos^{2} t + \sin^{2} t \, dt = \frac{1}{2} \int_{0}^{2\pi} dt = \pi.$$

Theorem 3.8 (Change of variables). Let $U \subset \mathbb{R}^n$ be an open set, and get $g: U \to \mathbb{R}^n$ be a bijective, continuously differentiable map such that $\det Dg(x) \neq 0$ on U. If $f: g(U) \to \mathbb{R}$ is integrable, then

$$\int_{q(U)} f = \int_{U} f \circ g \, |\det Dg|$$