



Syracuse University

MATH 701: Real Variables I

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0 Introduction

0.1 Course Description

MAT 701 Real Variables I: Measure and integration, including basic theorems on integration and differentiation of sequences of functions; modes of convergence, product measures.

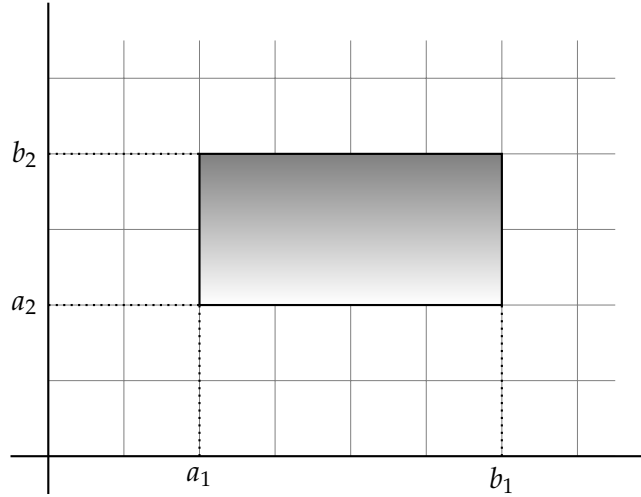
0.2 Disclaimer

These notes were taken in Fall 2018 in a course taught by Professor Leonid Kovalev. In some places, notation/material has been changed or added. Any errors in this text should be attributed to the typist – Caleb McWhorter – and not the instructor or any referenced text.

1 Outer Measure & Lebesgue Measure

1.1 Exterior Measure

We want to assign a notion of ‘size’ to sets. We denote this ‘size’ by ν . Let $\mathbb{R}^n = \{(x_1, \dots, x_n) : x_i \in \mathbb{R}\}$ denote ordinary Euclidean space. By an ‘interval’ in \mathbb{R}^n , we mean a set $\{(a_1, b_1) \times (a_2, b_2) \times \dots \times (a_n, b_n) : a_i, b_i \in \mathbb{R}^n, a_i < b_i\}$. By a closed interval in \mathbb{R}^n , we mean a set $\{[a_1, b_1] \times [a_2, b_2] \times \dots \times [a_n, b_n] : a_i, b_i \in \mathbb{R}^n, a_i < b_i\}$. We will often say ‘box’, which will always mean an open or closed interval in \mathbb{R}^n .



In defining a ‘size’ for sets, it makes sense to begin with a simple shape like a box. In the case of the plane, we know the a good notion of size is the area, and the area of a box $[a_1, b_1] \times [a_2, b_2]$ is $(b_1 - a_1) \cdot (b_2 - a_2)$. We can immediately generalize this to \mathbb{R}^n as follows: if $I \subset \mathbb{R}^n$ is an interval, then we define

$$\nu(I) := \prod_{j=1}^n (b_j - a_j) = (b_1 - a_1)(b_2 - a_2) \cdots (b_n - a_n)$$

But the question remains, how do we generalize this to an arbitrary region E ? Given the above definition, it is natural to try to generalize to arbitrary sets E by approximating E by boxes, i.e. an open covering $\{I_k\}$ of E by intervals (see Figure 1).

It then becomes clear that whatever the measure of E is, it should satisfy

$$\nu(E) \leq \sum_k \nu(I_k),$$

since the intervals cover E . After all, it would be strange indeed to allow E to have greater measure than its covering. Furthermore, if a notion of measure is going to be defined in

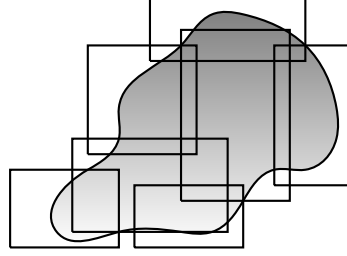


Figure 1: Approximating a set E by intervals.

terms of coverings, then the measure need be invariant of the choice of covering. These ideas are our guiding principles. So we take the following definition

Definition (Outer Measure). For a set $E \subset \mathbb{R}^n$, the outer measure (or exterior measure) of E , denoted $|E|_e$, is the function $\nu : \mathbb{R}^n \rightarrow [0, \infty)$ given by

$$|E|_e := \inf_{E \subset \bigcup_k I_k} \sum \nu(I_k),$$

where the infimum is taken over all *countable* coverings $\{I_k\}$ of E .

The coining of the term ‘outer measure’ is immediately obvious—we are measuring the size of a set via external objects, namely the open cover. However, less obvious is the need to restrict to countable coverings. The need to eliminate uncountable coverings is apparent, as then trouble arises defining the summation. But why not only allow finite coverings? For this consider the case $E = \mathbb{Q} \cap [0, 1]$. In any finite covering of $[0, 1]$ by intervals, the intervals cannot all be pairwise disjoint. The reader will confirm, with a bit of thought, that if the intervals were all pairwise disjoint then there would be a rational number missed by the ‘covering.’ But this contradicts the fact that the collection was an open cover. The only possible open covering by intervals is the entire interval itself so that $\inf \sum \nu(I_k) = 1$. This violates the notation that there isn’t any ‘length’ or ‘area’ here since we have a sparse collection of points.

Furthermore, the same logic applies to the set $E' = \mathbb{Q}^c \cap [0, 1]$. So if the outer measure of E were 1, then this would be true too of E' . But clearly the outer measure of $[0, 1]$ is 1. Now $[0, 1] = E \cup E'$, and $E \cap E' = \emptyset$. As $1 + 1 \neq 1$, this breaks countable subadditivity of the measure we are trying to define. By defining the outer measure in terms of countable covers, we obtain the expected answer $|E|_e = 0$.

Example 1.1. Let $E = \mathbb{Q} \cap [0, 1]$ and $\epsilon > 0$ be given. Since \mathbb{Q} is countable, so too is E countable. Enumerate the rationals in E as $\{q_1, q_2, \dots, q_n, \dots\}$. Now the set $\{O_n\}_{n \in \mathbb{N}}$, where $O_n := (q_n - \frac{1}{2^{n+k+1}}, q_n + \frac{1}{2^{n+k+1}})$ and $k \in \mathbb{N}$ is fixed, is a (countable) open covering of E by intervals. Furthermore, the O_n are pairwise disjoint. Choose k sufficiently large so

that $2^{-k} < \epsilon$. The measure of this covering is then

$$\sum_{n=1}^{\infty} \nu(O_n) = \sum_{n=1}^{\infty} \left[\left(q_n + \frac{1}{2^{n+k+1}} \right) - \left(q_n - \frac{1}{2^{n+k+1}} \right) \right] = \sum_{n=1}^{\infty} 2 \cdot \frac{1}{2^{n+k+1}} = \frac{1}{2^k} \sum_{n=1}^{\infty} \frac{1}{2^n} = \frac{1}{2^k} < \epsilon.$$

◁

It is also important to note that our notion of measure preserves our notion of ‘size’ for intervals. Before proving this in the general case, we begin with the case for intervals in \mathbb{R} .

Proposition 1.1. $|[a, b]|_e = b - a$.

Proof. Clearly, $|[a, b]|_e \leq b - a$, so it remains to show that $b - a \leq |[a, b]|_e$. Suppose that $[a, b]$ is a finite union of intervals of the form $[c_j, d_j]$. There exists j_1 such that $c_{j_1} \leq a$. If $d_{j_1} \geq b$, then $|[a, b]|_e \geq d_{j_1} - c_{j_1} \geq b - a$, and we are done. Otherwise, it must be that $d_{j_1} < b$. There then exists j_2 such that $c_{j_2} \leq d_{j_1}$. Continue this process inductively until one finally obtains $d_{j_r} \geq b$. But then taking the sum of these differences, one obtains a telescoping series

$$\underbrace{(d_{j_r} - c_{j_r})}_{\geq 0} + \underbrace{(d_{j_{r-1}} - c_{j_{r-1}})}_{\geq 0} + \cdots + \underbrace{(d_{j_1} - c_{j_1})}_{\geq 0} \geq b - a.$$

□

We now prove this for general intervals.

Proposition 1.2. For an interval I , $|I|_e = \nu(I)$.

Proof. As I serves as an open covering for itself, we know that $|I|_e \leq \nu(I)$. It remains to show that $\nu(I) \leq |I|_e$. Suppose $\{I_k\}_k$ is a countable open covering of I , and let $\epsilon > 0$. Every interval I is contained in the interior of a slightly larger interval I' , i.e. $I \subseteq \text{Int}(I')$, where $\nu(I') - \nu(I) < \epsilon$. Let I'_k denote an interval containing I_k in its interior with $\nu(I'_k) \leq (1 + \epsilon)\nu(I_k)$. We know that $I \subset \bigcup_k \text{Int}(I'_k)$. Now I is compact so that Heine Borel implies that there is an integer N such that $I \subset \bigcup_{k=1}^N I'_k$. We know $\nu(I) \leq \sum_{k=1}^N \nu(I'_k)$. But then

$$\nu(I) \leq (1 + \epsilon) \sum_{k=1}^N \nu(I_k) \leq (1 + \epsilon) \sum_{k=1}^{\infty} \nu(I_k).$$

It then follows that $\nu(I) \leq |I|_e$, completing the proof. □

More generally, $\left| \bigcup_{k=1}^N I_k \right|_e = \sum_{k=1}^N \nu(I_k)$, provided the I_k are non-overlapping intervals, i.e. $\text{Int}(I_k) \cap \text{Int}(I_j) = \emptyset$, see Figure 2. The proof of this is rather ugly—an exercise in making ‘obvious’ geometric facts obvious, and an exercise in bookkeeping—and we shall not concern ourselves with it.

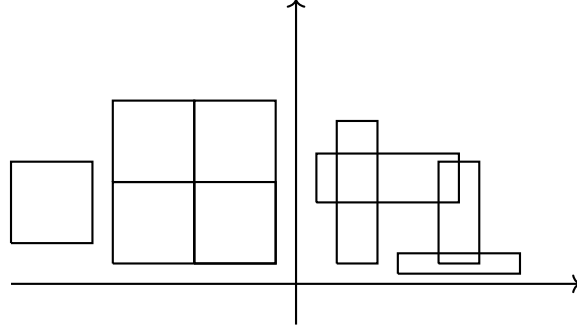


Figure 2: The intervals in Quadrant II are non-overlapping, while those in Quadrant I overlap.

As a final remark, observe that the exterior measure is a function to the nonnegative extended real line; that is, the exterior measure allows infinite values. Sets with an infinite exterior measure are considered measurable. For example, our intuition is that \mathbb{R} should have infinite length, and the reader routinely verifies that $|\mathbb{R}|_e = \infty$. The outer measure $|\cdot|_e$ defined above does meet all our guiding principles as the following proposition verifies.

Proposition 1.3.

(i) $\nu(\emptyset) = 0$.

(ii) *Monotonicity:* if $E_1 \subset E_2$, then $|E_1|_e \leq |E_2|_e$.

(iii) *Countable Subadditivity:*
$$\left| \bigcup_{k=1}^{\infty} E_k \right|_e \leq \sum_{k=1}^{\infty} |E_k|_e$$

Proof.

(i) This holds essentially by fiat.

(ii) This follows immediately from the fact that we are taking an infimum, and any open cover of E_2 is an open cover of E_1 .

(iii) Let $E := \left| \bigcup_{k=1}^{\infty} E_k \right|_e$. If any of the E_k have infinite exterior measure, the result is immediate. Assume then that $|E_k|_e < \infty$ for all k . Choose $\epsilon > 0$ and cover each E_k by intervals $\{I_n\}$ such that $\sum_n \nu(I_n) \leq |E_k|_e + \epsilon/2^k$. Then $E \subset \bigcup_{k,n} I_{k,n}$ and $|E|_e \leq \sum_{k,n} \nu(I_{k,n}) = \sum_k \sum_n \nu(I_{k,n})$. But then

$$|E|_e \leq \sum_k \left(|E_k|_e + \epsilon/2^k \right) = \epsilon + \sum_{k=1}^{\infty} |E_k|_e.$$

The result then follows by letting ϵ tend to 0. \square

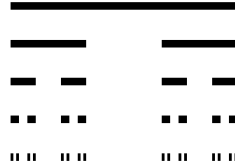
If one wants to generalize the notion of outer measures to spaces beyond \mathbb{R}^n , one can take the properties of Proposition 1.3 as the axioms for this abstract measure. Now we immediately have the following corollary:

Corollary 1.1. *Any subset of a set with outer measure zero has outer measure zero, and the countable union of sets with outer measure zero has outer measure zero. In particular, any countable set has outer measure zero.*

Proof. This follows immediately from the Countable Subadditivity of ν . \square

Now while Corollary 1.1 states that countable sets have outer measure zero, it need not be the case that uncountable sets need have positive measure.

Example 1.2 (Cantor Set). Begin with the closed unit interval $C_0 := [0, 1]$. From this interval, remove the middle third, i.e. $(\frac{1}{3}, \frac{2}{3})$, and label $C_1 := [0, \frac{1}{3}] \cup [\frac{2}{3}, 1]$. Inductively construct C_n by removing the middle third of each closed subinterval of C_{n-1} . We define the Cantor Set by defining $C := \lim_{n \rightarrow \infty} C_n$. The first few stages of the construction of C are shown below. The fact that the Cantor set is uncountable follows from the fact that it is a nonempty compact set without isolated points.



What is $|C|_e$? Note that C_n is the union of 2^n intervals, each having length $1/3^n$, and that $C \subset C_n$ for all $n \in \mathbb{N}$. But then the Cantor set is contained in a union of 2^n intervals with length $1/3^n$, which has total length $2^n \cdot 1/3^n = (2/3)^n$. The fact that $|C|_e = 0$ is then clear as $\lim_{n \rightarrow \infty} (2/3)^n = 0$. \triangleleft

Note in general we do not have $|\bigcup E_k|_e = \sum |E_k|_e$ even if the E_k are disjoint or even in the case of finite unions! Equality holds when the sets are, in a sense, ‘unentangled.’ By this, we mean that open coverings of one set tend to be disjoint from open coverings of the other set. If this is the case, the sets have to be covered separately, see the example on the left in Figure 3. However if the sets are ‘entangled’, then their open covers result in a great deal of ‘multiple-covering’ for the union. This excess covering allows one to more ‘efficiently’ cover the union—hence the smaller measure, see the example on the right in Figure 3.

There are many connections between Topology and Measure Theory, especially in the case of \mathbb{R}^n . Topology on \mathbb{R}^n is primarily interested in the structure of open and compact sets. This will prove useful for us since our notion of measure is defined in terms of open sets. As an example, take the following theorem.

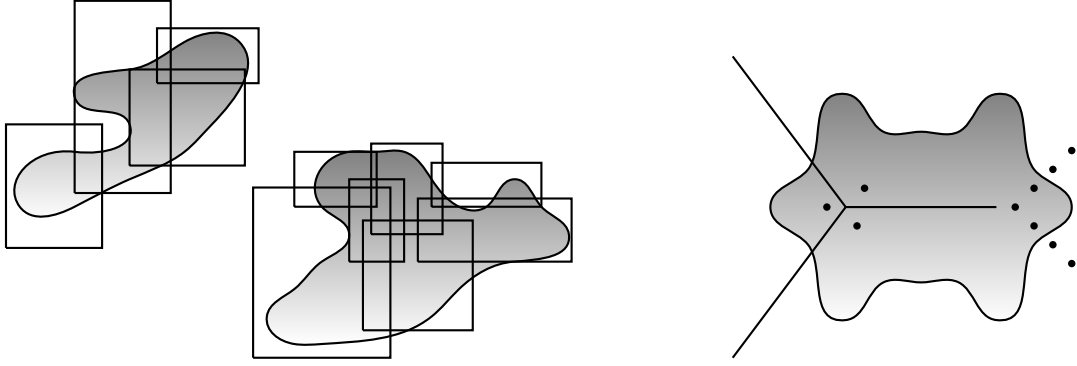


Figure 3: On the left, the two sets are ‘unentangled’ so that their opening coverings can be taken disjointly. On the right, the sets are the ‘blob’ and the line segments along with the points. These sets are ‘entangled’ and their open coverings will have ‘overlap.’

Theorem 1.1. *For all $E \subset \mathbb{R}^n$ and $\epsilon > 0$, there exists an open set G such that $E \subset G$ and $|G|_e \leq |E|_e + \epsilon$.*

Proof. Every interval I is contained in the interior of a slightly larger interval I' , i.e. $I \subseteq \text{Int}(I')$, where $\nu(I') - \nu(I) < \epsilon$. Choose intervals I_k such that $E \subset \bigcup_{k=1}^{\infty} I_k$ and $\sum_k \nu(I_k) \leq |E|_e + \epsilon/2$. Now find I'_k such that $I_k \subset \text{Int}(I'_k)$ and $\nu(I'_k) < \nu(I_k) + \epsilon/2^{k+1}$. Let $G = \bigcup_k \text{Int}(I'_k)$. By construction, G is an open set containing E . To complete proof, observe

$$|G|_e \leq \sum_{k=1}^{\infty} \nu(I'_k) \leq \sum_{k=1}^{\infty} \nu(I_k) + \epsilon \sum_{k=1}^{\infty} \frac{1}{2^{k+1}} \leq |E|_e + \epsilon.$$

□

Corollary 1.2. *For every set E , there exists a G_δ set G such that $E \subset G$ and $|E|_e = |G|_e$.*

Proof. By Theorem 1.1, we know that for every $k \in \mathbb{N}$, there is an open set $G_k \supset E$ with $|G_k|_e \leq |E|_e + 1/k$. Define $G := \bigcap_{k=1}^{\infty} G_k$. Then G is a G_δ set and $G \supset E$. Finally,

$$|E|_e \leq |G|_e \leq |G_k|_e \leq |E|_e + \frac{1}{k},$$

so that $|E|_e = |G|_e$.

□

The set G_δ referred to in Corollary 1.2 is a general type of set which will be of occasional interest and use for us:

Definition (G_δ -Set). A G_δ set is a countable intersection of open sets.

Notationally, the choice of ‘ G ’ is because the set is open, and the choice of ‘ δ ’ stems from the fact we are using an intersection.

Theorem 1.1 says we can always approximate any set by an open set with approximately the same size, i.e. approximately the same exterior measure. However, this does not mean that $|G \setminus E|_e \leq \epsilon$. We do know that $G = E \cup (G \setminus E)$. By subadditivity, we have $|G|_e \leq |E|_e + |G \setminus E|_e$. But we do not know the measure of the second set. In fact, it is possible that $|G \setminus E|_e$ could be very large. Now as it turns out for compact sets, it is sufficient to consider only finite coverings.

Proposition 1.4. *If $K \subset \mathbb{R}^n$ is compact, then $|K|_e = \inf \{\sum_k \nu(I_k) : \{I_k\} \text{ finite cover of } K\}$.*

Proof. Let $\epsilon > 0$. By Theorem 1.1, every interval is contained some set $\text{Int}(I')$, where $\nu(I') < \nu(I) + \epsilon$. Given a countable cover I_k of K , choose I'_k such that $I_k \subset \text{Int}(I'_k)$ and $\nu(I'_k) < \nu(I_k) + \epsilon/2^k$. Then $\{\text{Int}(I'_k)\}_k$ is an open covering for K , so there exists a finite subcovering $\{I_{k,n}\}_{n=1}^N$. Therefore, we have $K \subset \bigcup_j I'_{k,j}$. By countable subadditivity,

$$\sum_{n=1}^N \nu(I'_{k,n}) < \epsilon + \sum_{n=1}^N \nu(I_k),$$

as desired. □

1.2 The Lebesgue Measure

We now can define our notion of measurability with our notions of exterior measure firmly in place.

Definition (Measurable). A set $E \subset \mathbb{R}^n$ is (Lebesgue) measurable if for all $\epsilon > 0$, there exists an open set G such that $G \supset E$ and $|G \setminus E|_e < \epsilon$.

Essentially, a set is measurable if it can be well approximated by open sets. We choose the above notion of ‘closeness’ in order to obtain additivity of measures. Notice we also have mentioned the underlying topology via the use of ‘open.’ We are able to avoid invoking the underlying topology using greater abstraction, which shall come later. Note that we always have an open set such that $|G|_e < |E|_e + \epsilon$ (c.f. Theorem 1.1), but this alone is weaker than the above definition; that is, if E is measurable then it satisfies the properties in Theorem 1.1. As a matter of notation, if E is measurable, we define $|E| := |E|_e$. The following propositions follow immediately from our definition.

Proposition 1.5. *Every open set is measurable.*

Proof. If E is an open set, choose $G = E$. □

Proposition 1.6. *If $|E|_e = 0$, then E is measurable.*

Proof. Choose G such that $|G|_e < |E|_e + \epsilon = \epsilon$. But then $|G \setminus E|_e \leq |G|_e < \epsilon$. \square

Proposition 1.7. *A countable union of measurable sets is measurable, and*

$$|E| \leq \sum_k |E_k|.$$

Proof. Let $\epsilon > 0$. For each k , choose an open set G_k such that $E_k \subset G_k$ and $|G_k \setminus E_k|_e < \epsilon/2^k$. Now $G := \bigcup_k G_k$ is open, and $E \subset G$. Moreover since $G \setminus E \subset \bigcup_k (G_k \setminus E_k)$, we have

$$|G \setminus E|_e \leq \left| \bigcup_k (G_k \setminus E_k) \right|_e \leq \sum_k |G_k \setminus E_k|_e < \epsilon.$$

Therefore, $\bigcup_k E_k$ is measurable. The fact that $|\bigcup_k E_k| \leq \sum_k |E_k|$ follows from Proposition 1.3. \square

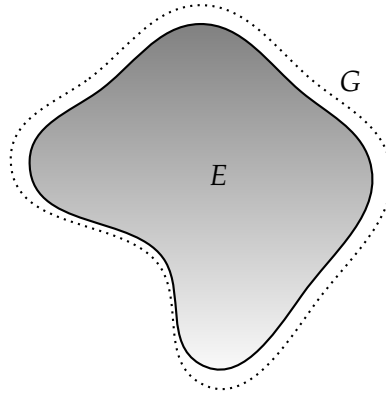


Figure 4: Open sets are not ‘fuzzy’ since they have a good notion of boundary. Similarly since G is not fuzzy and approximates E , it must be that E is not too close to being ‘fuzzy.’

Proposition 1.8. *All intervals are measurable.*

Proof (Sketch). We prove this only in the two dimensional case to avoid unnecessary complications. Given an interval $I = [a, b]$, choose $I' = [a', b']$ such that $I \subset \text{Int } I'$ and $\nu(I') < \nu(I) + \epsilon$. We need to show that $|I' \setminus I|_e < \epsilon$. Now $I' \setminus I = [a', a] \cup (b, b']$, and $|I' \setminus I|_e \leq a - a' + b' - b = (b' - a') - (b - a) < \epsilon$, as desired. Alternatively, one can use the fact that I is the union of its interior and its boundary, and that the boundary has

measure zero. This alternative proof is left to the reader. \square

As we have seen, open sets are easily seen to be measurable. But the case of closed sets is more complicated. For example in \mathbb{R} , every open set is the countable union of intervals of the form (a_k, b_k) . These intervals can even be taken to be disjoint. However, the same is not true for closed sets—take the Cantor set for example, c.f. Example 1.2. We do have that in \mathbb{R}^n , every open set is a countable union of non-overlapping intervals. To prove this we shall make use of dyadic cubes.

A dyadic cube of generation zero, \mathcal{D}_0 , are cubes with unit side lengths and integer vertices, i.e. $\mathcal{D}_0 := \{[0, 1]^n + \tau : \text{fixed } \tau \in \mathbb{Z}^n\}$. A generation one dyadic cube is $\mathcal{D}_1 := \{\frac{1}{2}Q : Q \in \mathcal{D}_0\}$. Generally, $\mathcal{D}_n := \{\frac{1}{2^n}Q : Q \in \mathcal{D}_0\} = \{\frac{1}{2^n}Q : Q \in \mathcal{D}_0\}$. Given \mathcal{D}_n , we say that \mathcal{D}_{n-1} is a parent of \mathcal{D}_n , and \mathcal{D}_i , where $i < n$, is an ancestor of \mathcal{D}_n . We say also that \mathcal{D}_{n+1} is a child of \mathcal{D}_n , and \mathcal{D}_j , where $j > n$, is a descendant of \mathcal{D}_n . One can allow n to be negative to create larger dyadic cubes. Define $\mathcal{D} := \bigcup_{k=0}^{\infty} \mathcal{D}_k$. If Q_1, Q_2 are dyadic, then either $Q_1 \subset Q_2$, $Q_2 \subset Q_1$, or they do not overlap. We now are in a position to prove the following lemma.

Lemma 1.1. *Every open set in \mathbb{R}^n is a countable union of non-overlapping intervals.*

Proof. Given an open set G , let $\{I_k\}$ be all dyadic cubes that are contained in G and for which their parent is not contained in G . By the selection of the I 's, it follows that they are pairwise disjoint for if I_k and I_j overlap, then one contains the other, contradicting the selection process. Clearly, we have selected only countably many intervals. Now if $x \in G$, there exists $r > 0$ such that there is an r -neighborhood of x contained in G . For sufficiently large n , all the cubes in \mathcal{D}_n have diameter less than r . But then there exists $Q \in \mathcal{D}_n$ such that $x \in Q$. Note that $Q \subset G$. Now either \mathcal{D}_n is contained in G or it has an ancestor that is contained in G . \square

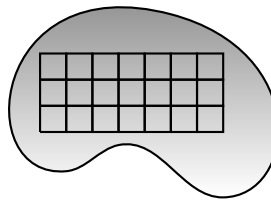


Figure 5: Tiling intervals inside an open set.

We can now make precise a discussion from earlier—if two sets are ‘unentangled’ then the measure of the union is the sum of the measures.

Lemma 1.2. *If $A, B \subset \mathbb{R}^n$ and $\text{dist}(A, B) > 0$, then $|A \cup B|_e = |A|_e + |B|_e$.*

Proof. We know by subadditivity that $|A \cup B|_e \leq |A|_e + |B|_e$. It remains to show that $|A|_e + |B|_e \leq |A \cup B|_e$. Let $\epsilon > 0$, and choose intervals $\{I_k\}$ such that $A \cup B \subset \bigcup_k I_k$ and $\sum_k |I_k| \leq |A \cup B|_e + \epsilon$. Possibly partitioning each I_k into a finite number of subintervals, we may assume that $\text{diam}(I_k) < \text{dist}(A, B)$. ‘Sort’ the set $\{I_k\}$ into two sets $\{I'_k\}$ and $\{I''_k\}$ which cover A and B , respectively. Then

$$|A|_e + |B|_e \leq \sum_k |I'_k| + \sum_k |I''_k| = \sum_k |I_k| \leq |A \cup B|_e + \epsilon.$$

Therefore, $|A|_e + |B|_e \leq |A \cup B|_e$, as desired. \square

Theorem 1.2. *Every closed set $A \subset \mathbb{R}^n$ is measurable.*

Proof. Given $\epsilon > 0$, we can choose $G \supset A$ such that $|G|_e < |A|_e + \epsilon$. Now $G \setminus A$ is open. By Lemma 1.1, we can write $G \setminus A = \bigcup_{k=1}^{\infty} I_k$, where $\{I_k\}$ is a collection of non-overlapping open intervals. We want to show that $\sum \nu(I_k) < \epsilon$, which will imply that $|G \setminus A|_e < \epsilon$. It suffices to show that $\sum_{k=1}^N \nu(I_k) < \epsilon$ for all N . Define $K = \bigcup_{k=1}^N I_k$. Since K is the finite union of compact sets in \mathbb{R}^n , K is compact. Furthermore, K is disjoint from A . As the set A is closed and K is compact, it must be that $\text{dist}(K, A) > 0$. It then follows from Lemma 1.2 that

$$|K \cup A|_e = |K|_e + |A|_e = \sum \nu(I_k) + |A|_e \leq |G|_e < |A|_e + \epsilon.$$

\square

As it turns out, the measurability of a set is equivalent to the measurability of its complement. This is useful in circumstances where one set is easier to handle than the other.

Theorem 1.3. *E is measurable if and only if E^C is measurable.*

Proof. Suppose that E is measurable. For $k \in \mathbb{N}$, choose an open set G_k such that $E \subset G_k$ and $|G_k \setminus E|_e < 1/k$. Now as G_k is open, G_k^C is closed, and hence measurable by Theorem 1.2. Let $G := \bigcup_k G_k^C$. Being the countable union of measurable sets, G is measurable, and $G \subset E^C$. Write $E^C = G \cup Z$, where $Z = E^C \setminus G$. Then $Z \subset E^C \setminus G_k^C = G_k \setminus E$, and therefore, $|Z|_e < 1/k$ for all k . Hence, $|Z|_e = 0$, and so Z is measurable. But then E^C is the union of measurable sets, and is thus measurable. Finally, taking E to be E^C and rerunning the argument mutatis mutandis proves that if E^C is measurable, then E is measurable. \square

Corollary 1.3. *If E_1, E_2 are measurable, then $E_1 \setminus E_2$ is measurable.*

Proof. We know that $E_1 \setminus E_2 = E_1 \cap E_2^C$. But E_2^C is measurable by Theorem 1.3. Therefore, $E_1 \setminus E_2$ is measurable, being the intersection of measurable sets. \square

Our definition is measurable is essentially that a set E is measurable if it can be ‘approximated well’ by open sets containing it. This is an exterior notion of measure. We can also define measurability instead by saying a set E is measurable if it can be ‘approximated well’ internally by closed sets, as the following proposition shows.

Proposition 1.9. *A set $E \subset \mathbb{R}^n$ is measurable if and only if given $\epsilon > 0$, there exists a closed set $F \subset E$ such that $|E \setminus F|_e < \epsilon$.*

Proof. E is measurable if and only if E^C is measurable, i.e. if and only if given $\epsilon > 0$, there exists an open set G with $E^C \subset G$ and $|G \setminus E^C|_e < \epsilon$. But such an open set G exists, noting that $G \setminus E^C = E \setminus F$, if and only if G^C is closed, $F \subset E$, and $|E \setminus F|_e < \epsilon$. \square

The previous theorems and propositions have shown that the complements of measurable sets are measurable, and countable unions and intersections of measurable sets are measurable. Since we shall make use of them in the future, note that if $\{E_k\}$ is a collection of measurable sets, then so are \limsup and \liminf :

$$\limsup E_k = \bigcap_{j=1}^{\infty} \bigcup_{k=j}^{\infty} E_k \quad \text{and} \quad \liminf E_k = \bigcup_{j=1}^{\infty} \bigcap_{k=j}^{\infty} E_k,$$

though we shall not prove this. Now all of these previous theorems and propositions essentially follow because we are working in the context of a σ -algebra.

Definition (σ -algebra). A nonempty collection of sets Σ is called a σ -algebra if it satisfies

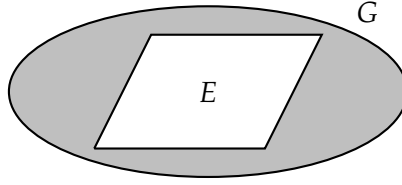
- (i) $E^C \in \Sigma$ whenever $E \in \Sigma$.
- (ii) $\bigcup_k E_k \in \Sigma$ whenever $E_k \in \Sigma$ for all k .

Note that any collection of sets closed under countable unions is closed under countable intersections as $[\bigcap_k U_k]^C = \bigcup_k U_k^C$. The empty set and entire space are necessarily measurable sets. Generally, they are elements of any σ -algebra. We shall be particularly interested in two specific σ -algebras:

Definition (Borel σ -algebra). The Borel σ -algebra is the smallest σ -algebra containing all open subsets of \mathbb{R}^n .

Definition (Lebesgue σ -algebra). The Lebesgue σ -algebra is the σ -algebra containing all measurable sets.

Note that the Borel σ -algebra is *strictly* contained in the Lebesgue σ -algebra—though this is not simple to see.



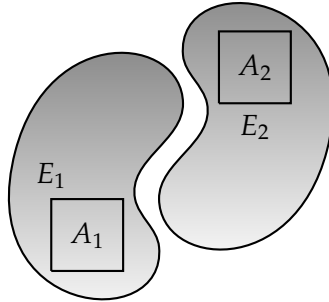
1.3 Properties of Measurable Sets

Our goal in this section is to examine measurable sets more closely. We begin by recalling that $|E| = |E|_e$ whenever E is measurable, and $|\bigcup_k E_k|_e \leq \sum_k |E_k|_e$, i.e. countable subadditivity.

Theorem 1.4 (Countable Subadditivity). *If $\{E_k\}$ is a collection of disjoint, measurable sets, then*

$$\left| \bigcup_{n=1}^{\infty} E_k \right| = \sum_{k=1}^{\infty} |E_k|.$$

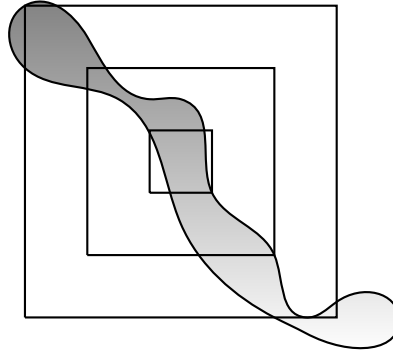
Proof. It is immediate that $|\bigcup_{n=1}^{\infty} E_k| \leq \sum_{k=1}^{\infty} |E_k|$, so that we need show $\sum_{k=1}^{\infty} |E_k| \leq |\bigcup_{n=1}^{\infty} E_k|$. Let $E := \bigcup_k E_k$. It suffices to show $\sum_{k=1}^N |E_k| \leq |E|$ for all N . If $\text{dist}(A, B) > 0$, we know that $|A \cup B|_e = |A|_e + |B|_e$. In particular, this holds if A, B are disjoint compact sets. First, assume that each E_k is bounded. For each $k \in \mathbb{N}$, choose a closed subset $A_k \subset E_k$



such that $|E_k \setminus A_k|_e < \epsilon/2^k$. Since $A_k \subset E_k$ and E_k is bounded, A_k is a closed, bounded set. Therefore, A_k is compact. We also have

$$\left| \bigcup_{k=1}^N A_k \right| = \sum_{k=1}^N |A_k| \geq \sum_{k=1}^N |E_k| - \epsilon.$$

But as $\bigcup_{k=1}^N A_k \subset E$, we have $|E| \geq \sum_{k=1}^N |E_k| - \epsilon$. Letting $\epsilon \rightarrow 0$ yields the desired inequality. This completes the proof in the case where each E_k is bounded.



Now suppose that some E_k may be unbounded. Let $I_j := [-j, j]^n$ for $j > 0$ and $Q_0 := \emptyset$. Define further $Q_j := I_j \setminus I_{j-1}$ for $j > 0$. Consider the sets $E_{k,j} := E_k \cap Q_j$ for $j, k = 1, 2, \dots$. Each of these sets are disjoint, bounded, and measurable. It is also routine to verify that

$$\begin{aligned} E_k &= \bigcup_j E_{k,j} = \bigcup_j (E_k \cap Q_j) \\ E &= \bigcup_k E_k = \bigcup_{k,j} E_{k,j} \end{aligned}$$

But then by the case above, we have

$$|E| = \left| \bigcup_k E_k \right| = \left| \bigcup_{k,j} E_{k,j} \right| = \sum_{k,j} |E_{k,j}| = \sum_k \sum_j |E_{k,j}| = \sum_k |E_k|.$$

□

Corollary 1.4. *If $\{I_k\}$ is a sequence of nonoverlapping intervals, then $|\bigcup_k I_k| = \sum_k |I_k|$.*

Proof. It is immediate that $|\bigcup_k I_k| \leq \sum |I_k|$. We need prove the opposite inequality. As the I_k are disjoint, we know $|\bigcup_k I_k| \geq |\bigcup_k \text{Int } I_k| = \sum_k |\text{Int } I_k| = \sum |I_k|$. But then $|\bigcup_k I_k| = \sum_k |I_k|$. □

Corollary 1.5. *Suppose E_1 and E_2 are measurable, $E_2 \subset E_1$, and $|E_2| < \infty$. Then $|E_1 \setminus E_2| = |E_1| - |E_2|$.*

Proof. We know $E_1 = E_2 \cup (E_1 \setminus E_2)$ and $|E_1| = |E_2| + |E_1 \setminus E_2|$. The result then follows as $|E_2| < \infty$. □

We end this section with an interaction between the Lebesgue measure and monotone sequences of sets.

Proposition 1.10.

- (i) If $E_1 \subset E_2 \subset \dots$ are measurable, then $|\bigcup_k E_k| = \lim_{k \rightarrow \infty} |E_k|$; that is, if $E_k \nearrow E$, then $\lim_{k \rightarrow \infty} |E_k| = |E|$.
- (ii) If $E_1 \supset E_2 \supset \dots$ are measurable, $|E_k| < \infty$ for some k , then $|\bigcap_{k=1}^{\infty} E_k| = \lim_{k \rightarrow \infty} |E_k|$; that is, if $E_k \searrow E$ and $|E_k| < \infty$ for some k , then $\lim_{k \rightarrow \infty} |E_k| = |E|$.

Proof.

- (i) Let $E = \bigcup E_k$. If $|E_k| = \infty$ for some k , then $\lim_{k \rightarrow \infty} |E_k|$ and $|E|$ are infinite, and the result follows. Assume then that $|E_k| < \infty$ for all k . We want to create a disjoint union. We shall express E as $E = E_1 \cup (E_2 \setminus E_1) \cup (E_3 \setminus E_2) \cup \dots \cup (E_k \setminus E_{k-1}) \cup \dots$. Define $F_k := E_k \setminus E_{k-1}$. Observe the sets F_k are measurable and pairwise disjoint. Then by Theorem 1.4,

$$|E| = \sum_{k=1}^{\infty} |F_k| = \lim_{n \rightarrow \infty} \sum_{k=1}^n |F_k| = \lim_{n \rightarrow \infty} \left| \bigcup_{k=1}^n F_k \right| = \lim_{n \rightarrow \infty} |E_n|.$$

- (ii) Let $E = \bigcap E_k$. Since considering tails of a sequence are sufficient, we may assume without loss of generality that $|E_1| < \infty$. We again want to create disjoint, measurable sets. Define $F_k := E_k \setminus E_{k+1}$. Observe that the sets F_k are measurable, pairwise disjoint, and that $E_1 = E \cup \bigcup F_k$. Then by Theorem 1.4,

$$|E_1| = |E| + \left| \bigcup F_k \right| = |E| + \sum_{k=1}^{\infty} |F_k| = |E| + \sum_{k=1}^{\infty} (|E_k| - |E_{k+1}|) = |E| + |E_1| - \lim_{k \rightarrow \infty} |E_k|.$$

But as $|E_1| < \infty$, we must then have $|E| = \lim_{k \rightarrow \infty} |E_k|$.

□

Note that the requirement that $|E_k| < \infty$ for some k in the second part of Proposition 1.10 is necessary. For example, take $E_k = \mathbb{R}^2 \setminus C_k$, where C_k is the circle centered at the origin with radius k . Then $|E_k| = \infty$ for all k , and $E_k \searrow \emptyset$. However, $\lim_{k \rightarrow \infty} |E_k| = \infty$, while $|\emptyset| = 0$.

1.4 Characterization of Measurable Sets

By definition, A is measurable if and only if for all $\epsilon > 0$, there exists an open set $G \supset A$ such that $|G \setminus A|_e < \epsilon$. But this makes use of the underlying topology. Can we avoid this?

After all, this ‘open set-ness’ is about continuity. Continuity is not even a requirement for integration, which is an end goal of ours. So we would like a description of measurability which avoids Topology entirely. Ironically, we shall first need more Topology.

Recall that G_δ is a set which is the countable intersection of open sets, F_σ is the countable union of closed sets, $G_{\delta,\sigma}$ are the countable intersection of closed sets.¹ One can even continue this to absurdity, i.e. $G_{\delta\sigma\delta\sigma}$. Furthermore, these are all subclasses of the Borel σ -algebra. One may wonder if all Borel created by these procedures, i.e. formed by countable union or intersections of open or closed sets? This turns out to not be the case, though it is difficult to construct a set in this fashion which is not a Borel set. However, it is the case that measurable sets are ‘close’ to Borel sets, in the sense they differ by a set of measure zero, as the next results shows.

Theorem 1.5. *E is measurable if and only if $E = H \setminus Z$ if and only if $E = F \cup Z$, where H is of type G_δ , F is of type F_σ , and $|Z| = 0$.*

Proof. It is sufficient to prove that each of the statements is equivalent to the measurability of E . If E can be represented as $H \setminus Z$ or $F \cup Z$, then E is measurable, as H , F , and Z are measurable. Now assume that E is measurable. For each k , choose an open set G_k such that $E \subset G_k$ and $|G_k \setminus E| < 1/k$. Define $H := \cap G_k$. Clearly, H is of type G_δ , $E \subset H$, and $H \setminus E \subset G_k \setminus E$ for each k . But then defining $Z = H \setminus E$, we see that $|Z| = 0$, as desired. Furthermore, if E is measurable, then so too is E^C . Now using the previous work, write $E^C = (\cap G_k) \setminus Z$, where the G_k are open sets and $|Z| = 0$. But then taking complements yields $E = (\cup G_k^C) \cup Z$. But as the sets G_k are open, the sets G_k^C are closed, implying $F := \cup G_k^C$ is of type F_σ . \square

Remark. This is different (in fact stronger) than $E \subset H$, where H is a G_δ set with the same outer measure, i.e. $|H|_e = |E|_e$.

Theorem 1.5 essentially states that measurable sets are the Borel sets ‘plus or minus’ the measure zero sets. Note that we are still making Borel sets, which invoke Topology, so we still have not met our goal. However, we are now set up to prove Carathéodory’s Theorem, which we can use to meet our goal.

Theorem 1.6 (Carathéodory). *E is measurable if and only if for all $A \subset \mathbb{R}^n$, $|A \cap E|_e + |A \setminus E|_e = |A|_e$.*

Proof. Suppose that E is measurable. Choose a G_δ set H such that $A \subset H$ and $|H|_e = |A|_e$. Since $H = (H \cap E) \cup (H \setminus E)$, and these are measurable sets, we have

$$|A|_e = |H| = |H \cap E| + |H \setminus E|.$$

¹Generally, G denotes open, F denotes closed, δ is for countable intersections, and σ is for countable unions.

But then $|A|_e = |H| = |H \cap E| + |H \setminus E| \geq |A \cap E|_e + |A \setminus E|_e$. By subadditivity, we obtain $|A|_e \leq |A \cap E|_e + |A \setminus E|_e$. Therefore, $|A|_e = |A \cap E|_e + |A \setminus E|_e$.

Now suppose that E satisfies the condition that $|A \cap E|_e + |A \setminus E|_e = |A|_e$ for all $A \subseteq \mathbb{R}^n$. Let A be a G_δ set such that $E \subset A$ and $|E|_e = |A|$. Then on the left side of the equation, we have $|E|_e + |A \setminus E|_e = |A| = |E|_e$. Hence, $|A \setminus E|_e = 0$, so E is G_δ minus a measure zero set. Notice that we have subtracted $|E|_e$; hence, we need consider the case where $|E|_e = \infty$.

Assume that $|E|_e = \infty$. We know that $E = \bigcup_{j=1}^{\infty} E_j$, where $E_j := E \cap [-j, j]^n$. So it suffices to prove that E_j has the property $|A \cap E_j|_e + |A \setminus E_j|_e = |A|_e$ for all $A \subseteq \mathbb{R}^n$ and $j \in \mathbb{N}$. [Since each E_j is measurable and the countable union of the E_j is measurable, E would then be measurable.] Certainly by subadditivity, $|A|_e \leq |A \cap E_j|_e + |A \cap E_j^C|_e$. It remains to show $|A|_e \geq |A \cap E_j|_e + |A \cap E_j^C|_e$. By assumption for all $A \subset \mathbb{R}^n$, $|A \cap E|_e + |A \setminus E|_e = |A|_e$. We know also that $Q_j := [-j, j]^n$ and Q_j^C satisfy this property, so $|A \cap E|_e = |(A \cap E) \cap Q_j|_e + |(A \cap E) \cap Q_j^C|_e$. Similarly, we have $|A \cap E^C|_e = |(A \cap E^C) \cap Q_j|_e + |(A \cap E^C) \cap Q_j^C|_e$. Combining these gives

$$\begin{aligned} |A|_e &= |A \cap (E \cap Q_j)|_e + |A \cap (E \cap Q_j^C)|_e + |A \cap (E^C \cap Q_j)|_e + |A \cap (E^C \cap Q_j^C)|_e \\ &= |A \cap E_j|_e + |A \cap E \cap Q_j^C|_e + |A \cap E^C \cap Q_j|_e + |A \cap E_j^C|_e \\ &\geq |A \cap E_j|_e + |A \cap E_j^C|_e, \end{aligned}$$

as desired. \square

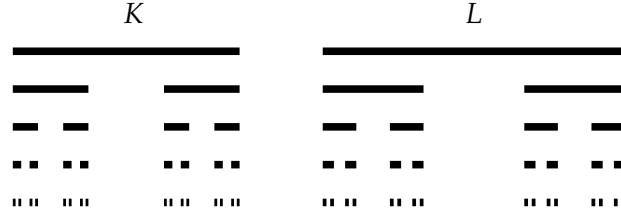
Note that in Theorem 1.6, there is no mention of openness or closed-ness. We can then take this as our definition of measurability, i.e. define measurable sets to be the sets which split every set (independent of measurability) into pieces which are additive with respect to the outer measure. This is the one of the standard approaches when defining measures abstractly. Thus, we have completed our goal of finding a way to avoid invoking the underlying topology.

1.5 Continuous & Lipschitz Transformations

Here we shall investigate the interaction between continuous functions and measurability. We begin with an important question: if $E \subset \mathbb{R}^n$ is measurable and $f : \mathbb{R}^n \rightarrow \mathbb{R}^n$ is continuous, is $f(E)$ always measurable? The general answer is no, but there are special cases where this does hold. For example, measures of sets are invariant under rotations of \mathbb{R}^n —though we will not prove this, see [WZ77, p. 36]. This also shows that the exterior measure (hence also the Lebesgue measure) is invariant under the choice of an orthogonal basis for \mathbb{R}^n . However, what other functions or transformations preserve measures? We first show that continuity alone will never be enough.

Lemma 1.3. *There is a continuous function $f : \mathbb{R} \rightarrow \mathbb{R}$ and a set E with $|E| = 0$, such that $|f(E)| > 0$.*

Proof (Sketch). We construct a ‘fat Cantor’ set by following the same procedure as in the Cantor set, but instead removing the middle fourths of each interval. At each stage, 2^j intervals of length 4^{-j} are removed. Denote the ordinary Cantor set by K and the fat Cantor set by L . We construct a continuous function $f : [0, 1] \rightarrow [0, 1]$ such that $f(K) = L$.



Map K_j , the j th step in the construction of K , onto L_j in a piecewise linear fashion, i.e. stretch each interval in K_j to the corresponding interval in L_j . Clearly, f_j is continuous and $\sup |f_j - f_{j+1}| \leq 2^{-j}$. But then $f_j \rightarrow f$ uniformly so that f is continuous and $f(K) = L$. We can extend f to \mathbb{R} by translation, i.e. define $f(x)$ for $x \notin [0, 1]$ via $f(x) = f(a)$, where $a = x + n$ and $n \in \mathbb{Z}$ is an integer such that $x + n \in [0, 1]$. Finally, recall that the ordinary Cantor set has measure 0, as was shown in Example 1.2, and it is routine to verify that $|L| > 0$. \square

Before answer the question poised at the beginning of this section, we need two quick results, assuming the existence of a nonmeasurable set, which we shall prove in the next section.

Lemma 1.4. *Let E be a measurable subset of \mathbb{R}^1 with $|E| > 0$. Then the set of differences $\{d : d = x - y, x \in E, y \in E\}$ contains an interval centered at the origin.*

Proof. See [WZ77, p. 46]. \square

Lemma 1.5. *Any set in \mathbb{R}^1 with positive outer measure contains a nonmeasurable set.*

Proof. Suppose A is a set with $|A|_e > 0$. Let E be the nonmeasurable Vitali set. For $q \in \mathbb{Q}$, let E_q denote the translation of E by q , i.e. $E_q = \{e + q : e \in E\}$. The sets E_q are disjoint, and $\cup E_q = \mathbb{R}$. But then it must be that $A = \cup (A \cap E_q)$. By Lemma 1.4, if $A \cap E_q$ is measurable then $|A \cap E_q| = 0$. But as $|A|_e > 0$, there must be some $q \in \mathbb{Q}$ such that $A \cap E_q$ is not measurable. \square

We now sketch the idea why it need not be the case that $f(E)$ is measurable even when E is measurable. Choose f and E as in Lemma 1.3. By Lemma 1.5, there exists a

non-measurable subset of $f(E)$, say W . Let $V := f^{-1}(W) \cap E$. Observe that $f(V) = W$, and that V is measurable (being a subset of measure zero). Therefore, the image of a measurable set under a continuous function need not be measurable, as claimed. The next obvious question is, if continuity is not sufficient to force the image of measurable sets to be measurable, is there a stronger condition that will? We recall a definition.

Definition (Lipschitz). We say that f is Lipschitz if there exists L such that $|f(x) - f(y)| \leq L|x - y|$ for all x, y in the domain of f .

Lipschitz functions are the special case of uniformly continuous functions where we can always choose $\delta = \epsilon/L$. Therefore, if f is Lipschitz then f is uniformly continuous, and hence also continuous. However, not all uniformly continuous functions are Lipschitz, and certainly not all continuous functions are Lipschitz. The Lipschitz condition can be thought of as the boundedness of secant lines for a given function. Hence functions which are differentiable with bounded derivative on an interval are Lipschitz.

Example 1.3. The function $f(x) = x^{1/3}$ is Lipschitz on any interval not containing the origin as a limit point since $f'(x) = \frac{1}{3\sqrt[3]{x^2}}$ is bounded on such intervals. Indeed, if I is an interval containing the origin and $x \in I$, then

$$\frac{f(x) - f(0)}{x - 0} = \frac{\sqrt[3]{x}}{x} = \frac{1}{\sqrt[3]{x^2}}.$$

Taking x 'close' to 0 shows that the slopes of secant lines to $f(x)$ are unbounded. One can use the same process to verify this for any set containing 0 as a limit point. \triangleleft

Example 1.4. The function $f : [0, 1] \rightarrow [0, 1]$ given by $f(x) = \sqrt{x}$ is uniformly continuous on $[0, 1]$, being a continuous function on a compact interval. However, f is not Lipschitz continuous. If f were Lipschitz with constant L , then we would have $|\sqrt{x} - \sqrt{y}| \leq L|x - y|$ for all $x, y \in [0, 1]$. Choose $x = 0$ and $y = \frac{1}{4k^2}$. We have $x, y \in [0, 1]$ but

$$|\sqrt{x} - \sqrt{y}| = \frac{1}{2K} > \frac{1}{4K} = K|x - y|.$$

\triangleleft

Unlike general continuous functions or the function in Lemma 1.3, Lipschitz maps preserve sets of measure zero.

Theorem 1.7. If $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is Lipschitz and $|E| = 0$, then $|f(E)| = 0$.

Proof. Recall that $|E|_\epsilon = \inf\{\sum \nu(I_k) : \bigcup I_k \supset E, \text{ and } I_k \text{ cubes}\}$. Since $|E| = 0$, for every $\epsilon > 0$, there exists a countable cover of E by intervals $\{I_k\}$ such that $\sum |I_k| < \epsilon$. Choose

this covering to consist of dyadic cubes. By Lipschitz continuity, $f(I_k)$ is contained in an interval of width at most Lw_k , where L is the Lipschitz constant of f and w_k is the maximum width of the cube. Then $|f(I_k)| \leq L^d |I_k|$. Therefore, we have

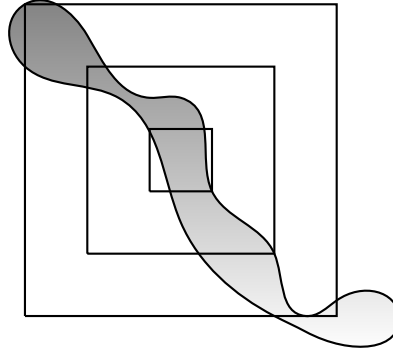
$$|f(E)| \leq L^d \sum |I_k| < L^d \epsilon.$$

The result then follows. \square

Furthermore, Lipschitz functions preserve measurability, though not necessarily the value of the measure.

Theorem 1.8. *If E is measurable and f is Lipschitz, then $f(E)$ is measurable.*

Proof. Since E is measurable, write $E = H \cup Z$, where H is F_σ and $|Z| = 0$. Now F_σ sets are the countable union of closed sets, which is equivalent to the countable union of compact sets since any union of closed sets can be written as a countable union of compact sets, i.e. $F = \cup (F \cap [-j, j]^n)$. Now f is Lipschitz, hence continuous, so that f sends compact sets to compact sets. Then it must be that f preserves ' F_σ -ness.' We know by Theorem 1.7, if $|E| = 0$ then $f(E) = 0$. But then $f(E) = f(H) \cup f(Z)$, a union of F_σ set and a set of measure zero, hence measurable. \square



The proof of Theorem 1.8 involves a common trick: although a whole function may not be Lipschitz, 'pieces' of it may be, and these could be brought together to finish a result. This invoked the same underlying idea as in the proof of Theorem 1.4. Though Theorem 1.8 showed that Lipschitz functions preserve measurability, they do not generally preserve the measure. So while we cannot know that the measure of $f(E)$ is, we know that $f(E)$ is indeed measurable. We finish by commenting that in the special case of Lipschitz functions which are linear maps of \mathbb{R}^n , we can actually compute the measure of the resulting image.

Theorem 1.9. *Let T be a linear transformation of \mathbb{R}^n , and let E be a measurable set. Then $|T(E)| = \delta |E|$, where $\delta = |\det T|$.*

Proof. Let $\delta = |\det T|$. It is routine to verify that $|T(I)| = \delta|I|$ for any interval I . Now let $E \subseteq \mathbb{R}^n$ be a measurable subset, and $\epsilon > 0$. Choose intervals $\{I_k\}$ covering E with $\sum |I_k| < |E|_e + \epsilon$. Then

$$|T(E)|_e \leq \sum |T(I_k)| = \delta \sum |I_k| < \delta(|E|_e + \epsilon).$$

But then $|T(E)|_e \leq \delta |E|_e$.

It remains to show that $\delta |E|_e \leq |T(E)|_e$. Choose an open set $G \supset E$ with $|G \setminus E| < \epsilon$. Write G as a union of nonoverlapping intervals $\{I'_k\}$. By the work above, we may assume that $\delta > 0$. Since $T(I_k)$ are also nonoverlapping intervals,

$$|T(G)| = \sum |T(I_k)| = \delta \sum |I_k| = \delta |G|.$$

But as $E \subset G$, we have $\delta |E| \leq \delta |G| = |T(G)|$. Moreover,

$$|T(G)| \leq |T(E)| + |T(G \setminus E)| \leq |T(E)| + \delta \epsilon.$$

Combining inequalities yields $\delta |E| \leq |T(E)| + \delta \epsilon$, so that $\delta |E| \leq |T(E)|$. \square

1.6 A Nonmeasurable Set

Notice that ‘all’ the sets we have encountered thus far were measurable. Indeed, most sets you ‘meet on the street’ will be measurable. Nonmeasurable sets are true pathologies, which do not even exist depending on your logical axioms.

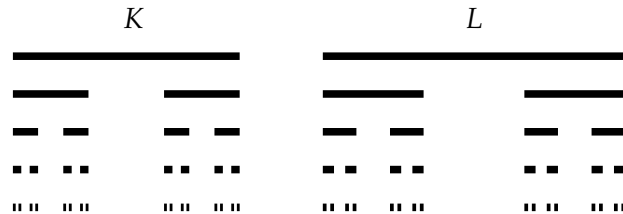
Recall the example from Lemma 1.3. We constructed a fat Cantor set, L . Note that removing fourths was not the only possible choice in the construction of L . Suppose at each stage we cut out lengths $\epsilon_1, \epsilon_2, \dots, \epsilon_n, \dots$. We want

$$\sum_{k=1}^{\infty} 2^{k-1} \epsilon_k = \frac{1}{2},$$

or any other number in $[0, 1)$ so that our final set L has positive measure. Clearly, choosing $\epsilon_k = \frac{1}{4^k}$ works. Furthermore,

$$|f_k - f_{k+1}| \leq 2^{-k}, |f_k^{-1} - f_{k+1}^{-1}| \leq 2^{-k}$$

so that f is a homeomorphism.



Now when $W \subset F$ is nonmeasurable, $V = f^{-1}(W)$ is measurable but not Borel. If V were Borel, then $W = (f^{-1})^{-1}(V)$ would also be Borel, hence measurable, as the preimage of Borel sets are Borel, c.f. Problem 8. However, it need not be the case that the continuous image of a Borel set is measurable.

Theorem 1.10. *The continuous image of a Borel set is measurable.*

Though we shall not prove this here, this was shown by Lusin and his student Souslin in 1916, after reading a paper of Lebesgue and noting many errors including asserting that projections of Borel sets were Borel. Souslin's construction is a bit detailed, though not long, and we shall not pursue it here.

As mentioned previously, non-measurable sets are rare in practice, though they certainly do exist. To this day, no explicit examples of non-measurable sets exist. Indeed, their very existence hinges on the Axiom of Choice.

Axiom of Choice: For all collections of nonempty sets $\{E_\alpha\}$, one can choose an element from each E_α , i.e. there exists a function $f : \mathcal{A} \rightarrow \bigcup E_\alpha$ such that $f(\alpha) \in E_\alpha$ for all α . Said differently, $\prod_{\alpha \in \mathcal{A}} E_\alpha \neq \emptyset$.

The Axiom of Choice was once a point of great contention among mathematicians at the turn of the century as many counterintuitive results follow from its assumption. However generally today, the axiom is assumed without grudge. We shall assume it here. With this final ingredient, we can construct a nonmeasurable set.

Theorem 1.11 (Vitali Set). *There exists a nonmeasurable set.*

Proof. Define an equivalence relation on \mathbb{R} via $x \sim y$ if $x - y \in \mathbb{Q}$. Let $\{E_\alpha\}_{\alpha \in \mathcal{A}}$ be the set of equivalence classes. It is clear that the equivalence classes can be expressed $E_x := \{x + q : q \in \mathbb{Q}\}$. Furthermore, E_x and E_y are either identical or disjoint. There is one equivalence class consisting of all rational numbers, while the other equivalence classes consist entirely of irrational numbers, i.e. an irrational number translated by all possible rational numbers. Moreover since we can choose a representative of each equivalence class and all other elements of the class are the set of rational translates of this representative, each equivalence class is countable. However, the set of equivalence classes must be uncountable since their union is \mathbb{R} and each class is countable.

Consider the sets $\{\mathcal{A}_\alpha\}_\alpha$, where $\mathcal{A}_\alpha := E_\alpha \cap [0, 1]$. By the Axiom of Choice, we can choose an element from each set \mathcal{A}_α . Let $V \subset [0, 1]$ be the set of all the chosen elements. We claim that V is nonmeasurable. Now given two distinct points $x, y \in V$, we know that $x - y \notin \mathbb{Q}$, for otherwise they would come from the same equivalence class and we have chosen unique elements from each set. Therefore, every pair of points from V differ by an irrational number. Now $\mathbb{R} = V + \mathbb{Q} = \{v + q : v \in V, q \in \mathbb{Q}\}$, so that $\bigcup_{q \in \mathbb{Q}} (V + q) = \mathbb{R}$. If $|V|_e = 0$, then so too is $|V + q|_e = 0$. This would then imply that $|\mathbb{R}|_e = 0$, a contradiction.

Therefore, $|V|_e > 0$. Now observe $\bigcup_{q \in (\mathbb{Q} \cap [0,1])} (V + q) \subset [0,2]$. Now as the sets $V + q$ are distinct, if V were measurable, so too would $V + q$ be measurable, and we would have $|[0,2]| \geq \sum_{q \in (\mathbb{Q} \cap [0,1])} |V + q| = \infty$, a contradiction. \square

Notice that the sets $V + q$, if they were measurable, would necessarily have positive measure by the proof above. The proof then shows, in some sense, that these sets are so intertwined that one can fit infinitely many of them in an interval without having their measure go to zero.

Now we have stated that nonmeasurable sets are rare in practice. The next result shows that they are not, in principle, so rare. In fact, every measurable subset of \mathbb{R} with positive measure contains a nonmeasurable set.

Lemma 1.6. *If $E \subset \mathbb{R}$ and $|E| > 0$, then there exists an interval I such that $|E \cap I| > (1 - \epsilon)|I|$.*

Proof. There exists an open set G such that $G \supset E$ and $|G| < (1 + \epsilon)|E|$. Express G as the union of disjoint, open intervals, i.e. $G = \bigcup (a_k, b_k)$. For each k , let $E_k := E \cap (a_k, b_k)$, and note that each E_k are measurable and $E \subset \bigcup E_k$. Therefore, we have $|E| = \sum_k |E_k|$ and $|G| = \sum_k |(a_k, b_k)|$. If it were the case that for each k , $(1 + \epsilon)|E_k| \leq b_k - a_k$, then we would have

$$(1 + \epsilon)|E| = (1 + \epsilon) \sum |E_k| = \sum (1 + \epsilon)|E_k| \leq \sum (b_k - a_k) = \sum |(a_k, b_k)| = |G|,$$

a contradiction. Therefore, there is some k_0 such that $b_{k_0} - a_{k_0} < (1 + \epsilon)|E_{k_0}| = |E \cap (a_{k_0}, b_{k_0})|$. Defining $I := (a_{k_0}, b_{k_0})$ completes the proof. \square

Lemma 1.7. *If $E \subset \mathbb{R}$ and $|E| > 0$, then $E \setminus E$ contains a neighborhood of 0.*

Proof. We claim there exists $\delta > 0$ such that if $|x| < \delta$, then $E \cap (E + x) \neq \emptyset$. $x \in E \setminus E$ if and only if $E + x$ intersects E . $X = a - b$ if and only if $h + x = a$, where $a, b \in E$. Let I be an interval, $|E \cap I| > \frac{2}{3}|I|$. Let $\delta = \frac{1}{3}|I|$. If $|x| < \delta$.

$$|I \cup (I + \delta)| \leq \frac{4}{3}|I|.$$

If E and $E + x$ were disjoint, then $|(E \cap I) \cup (E \cap I + x)| = 2|E \cap I| > \frac{4}{3}|I| > |I \cup (I + x)|$, a contradiction (since one is a subset of the other). \square

Theorem 1.12. *If $E \subset \mathbb{R}$, $|E| > 0$, then there exists $E \subset E$ such that W is nonmeasurable.*

Proof. Since $\bigcup_{q \in \mathbb{Q}} (V + q) = \mathbb{R}$, where V is a Vitali set. We have $\bigcup_{q \in \mathbb{Q}} [(V + q) \cap E] = E$. So there exists q such that $|(V + q) \cap E|_e > 0$. But $(V + q) \cap E$ is nonmeasurable as $V + q$ has no measurable subsets of positive measure.

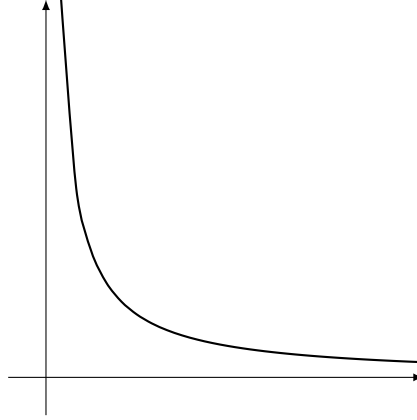
$((V + q) - (V + q)) = V - V$ is disjoint from \mathbb{Q} . \square

2 Lebesgue Measurable Functions

2.1 Measurable Functions

It is still our goal to build a generalized theory of integration for functions. However, there are two clear potential problems for functions which we might integrate:

- The graph of f could cover an infinite area and thus not be integrable in the traditional sense, e.g. $f(x) = \frac{1}{x}$ on the interval $(0, 1)$.



- The graph of f could be so irregular that the 'area underneath' is not well defined enough so that such a question even makes sense, e.g.

$$\chi_V(x) = \begin{cases} 1, & x \in V \\ 0, & x \notin V \end{cases}$$

where V is the Vitale set.

As in the case of the Riemann integral, only certain classes of functions will be integrable in the Lebesgue sense. It will be the goal of this section to define 'nice' classes of functions which address the issues above and which will later form a 'nice' class of functions with which to form an integration theory around.

Definition (Measurable Function). Let $E \subseteq \mathbb{R}^n$ and $\overline{\mathbb{R}}$ denote the extended real line. We say that a function $f : E \rightarrow \overline{\mathbb{R}}$ is (Lebesgue) measurable if for all $a \in \mathbb{R}$, the set $\{f > a\} := \{x \in E : f(x) > a\}$ is measurable.

Note that we define $\{f = \infty\} := \bigcap_{k \in \mathbb{Z}} \{f > k\}$ and $\{f = -\infty\} := \bigcap_{k \in \mathbb{Z}} \{-f > k\}$. The reason for this choice of definition is that the sets $\{f > a\}$ describe the distribution for the values of f . In some sense, the 'smoother' f is, the smaller the variety of such sets one

obtains from f . For example, when f is continuous, the sets $\{f > a\}$ are open sets since $f^{-1}((a, \infty))$ is an open set.

Furthermore, the measurability of a set should be linked in some way with the potential measurability of the functions on it and vice versa. If the set E is 'wild', less functions should be measurable on it. In fact, the measurability of E is equivalent to that of the set $\{f = -\infty\}$ since

$$E = \{f = -\infty\} \cup \left(\bigcup_{k=1}^{\infty} \{f > -k\} \right),$$

showing that if f is measurable, then E is measurable if and only if $\{f = -\infty\}$ is measurable.

Note that one could define Borel measurability similarly: $f : E \rightarrow \overline{\mathbb{R}}$ is Borel measurable, if for all $a \in \mathbb{R}$, the set $\{f > a\}$ is Borel. Thus, every Borel function is also measurable. We can of course define measurability of functions more generally: if M is a σ -algebra on \mathbb{R}^n , f is M -measurable if $\{f > a\} \in M$ for all $a \in \mathbb{R}$. For us, M will normally be the Lebesgue σ -algebra. Though what we do is in \mathbb{R}^n with the Lebesgue σ -algebra, much of what we do holds, or has parallel results, in other spaces and σ -algebras.

Example 2.1. If M is the σ -algebra generated by $[0, 1]$ and $[1, 2]$. What functions f are M -measurable?

M is σ algebra generated by $[0, 1]$ and $[1, 2]$. What f are M -measurable?

f over $[0, 1]$, $[1, 2]$ constant functions, over complements. Not M -measurable, $\chi_{\{3\}}$, $f(x) = x$ or $f(x) = \sin x$.

Before continuing, we make an important definition since the following abbreviation will serve as a constant shortening of the words needed to state, prove, and discuss results later, so we define it now.

Definition (Almost Everywhere, a.e.). We say that a property holds 'almost everywhere' if it holds everywhere except on a set of exceptional values with measure zero, e.g. f is finite almost everywhere signifies $|\{f = \infty\}| = 0$. We shall abbreviate this as 'a.e.'.

2.2 Properties of Measurable Functions

1. Continuous then measurable as $\{f > a\}$ are open.

2. Enough to check that $\{f > a\}$ is measurable for a in some dense set $\mathcal{D} \subset \mathbb{R}$. Why? For all real a , for all k , find $d_k \in \mathcal{D}$, $a < d_k < a + 1/k$. $\{f > a\} \cup_k \{f > d_k\}$.

3. Instead of $\{f > a\}$, it could be $\{f < a\}$ as $\{f \geq a\}$, $\{f \leq a\}$, can use any of these 4 ways to check measurability.

$$\begin{cases} \{f \geq a\} &= \bigcap_{k \in \mathbb{N}} \{f > a - 1/k\} \\ \{f > a\} &= \bigcup_{k \in \mathbb{N}} \{f \geq a + 1/k\} \end{cases}$$

Note that 2,3 are independent of the σ -algebra.

4. If f is measurable and $f = g$ a.e., then g is measurable (Borel fails this). Let $Z = \{f \neq g\}$, $|Z| = 0$. $\{f > a\} \setminus Z \subset \{g > a\} \subset \{f > a\} \cup Z$ but $\{f > a\} \setminus Z$ and $\{f > a\} \cup Z$ have the same measure.

Ex: $1/x$ is measurable on \mathbb{R} . Defined a.e. is okay.

5. iff f , then $f^{-1}(G)$ is measurable for all open $G \subset \mathbb{R}$.

(5b) also for all borel sets $G \subset \mathbb{R}$.

Proof. $G = \cup(a_k, b_k)$ so $f^{-1}(G) = \cup_k \{a_k < f < b_k\}$, measurable.

Proof. (5b) a homework generalization.

For all σ -algebra M , $\{E: f^{-1}(E) \in M\}$, where M Lebesgue σ -algebra, like M_f . By 5, M_f contains open sets, so it must contain Borel sets.

(5c) if f is finite a.e. and $f^{-1}(G)$ is measurable for all open $G \subset \mathbb{R}$, $\{f > a\} = f^{-1}((a, \infty))$.

Warning: the composition of measurable functions is not always measurable.

Recall C cantor set, F fat cantor set, $f(C) = F$, f homeomorphism $\mathbb{R} \rightarrow \mathbb{R}$. Let $W \subset F$ be nonmeasurable. Write χ_W as a composition $\chi_{g(W)} \circ g$

$\chi_{g(W)}$ is measurable, g is continuous (so measurable).

4. is what we gain for going Borel \rightarrow Lebesgue and this is the price we pay.

Rem: if f, g are Borel measurable, then $f \circ g$ is Borel measurable if for all $B \subset \mathbb{R}$ Borel we have $f^{-1}(B)$ is Borel, then $(f \circ g)^{-1}(B) = g^{-1}(f^{-1}(B))$, each composition on right Borel.

is \ (*)	Open	Borel	Measurable
Open	Continuous	—	—
Borel	Borel measurable	Borel measurable	—
Measurable	Measurable	Measurable	?

When can we say that $f \circ g$ is measurable? What data about f and g would ensure this? We know that if f is measurable and g is continuous does not work from previous discussion. One possibility ensuring this would be f is continuous and g is measurable as $(f \circ g)^{-1}(U) = g^{-1}(f^{-1}(U))$, as U is open $f^{-1}(U)$ is open and g^{-1} of open is measurable (assuming g takes values in \mathbb{R} and f is continuous on some open set containing the range of g). More generally, this will hold if f is Borel measurable on some Borel set containing the range of g and g measurable and real-valued.

Example 2.2. If f is measurable (and \mathbb{R} -valued) then f^2 is measurable, $f^2 = \phi \circ f$, where $\phi(t) = t^2$. Similarly, $|f|$ is measurable since $|f| = \phi \circ f$, where $\phi(t) = |t|$. Also, $\text{sgn}(f) = \begin{cases} 1, & f = 0 \\ 0, & f = 0 \\ -1, & f < 0 \end{cases}$ is measurable since sign is Borel measurable (simple to see by checking

preimages). So is $\text{floor}(x)$, where floor is supremum of integers which are at most x . As is $\text{ceilin}(x)$, defined mutatis mutandis. [They are both monotone,]

Theorem 2.1. *If f, g are measurable, then $f \pm g, fg$, and f/g ($g \neq 0$ a.e.) are measurable.*

Proof. $\{f + g > a\} = \bigcup_{b+c>a, b,c \in \mathbb{Q}} [\{f > b\} \cap \{g > c\}]$ hence measurable set (countable union of measurable sets). Just need to prove equality. \supset is obvious. Need only prove \subset . Suppose $f(x) + g(x) > a$. Let $\epsilon = f(x) + g(x) - a$. Use density of rationals to find $b \in \mathbb{Q}$ such that $f(x) - \epsilon/2 < b < f(x)$ and $c \in \mathbb{Q}$ such that $g(x) - \epsilon/2 < c < g(x)$. But then $b + c > f(x) + g(x) - \epsilon = a$.

Now $fg = \frac{1}{4} [(f+g)^2 - (f-g)^2]$ is then measurable. For quotient only need reciprocal. We know $\frac{1}{g}$ is $\phi \circ g$, where $\phi(t) = \frac{1}{t}$, which is continuous on $\mathbb{R} \setminus \{0\}$. Now g is nonzero almost everywhere, ϕ contains range of g (ignoring set where range is $g = 0$, since we can ignore for measurability). \square

Another way could combine is max of functions. Better to generalize to countable. So countable sup/inf of measurable functions is measurable.

Theorem 2.2. *The above statement, including infinite values.*

Proof. IF $f_k : E \rightarrow \overline{\mathbb{R}}$ is measurable for all $k \in \mathbb{N}$, then $\sup_k f_k$ and $\inf_k f_k$ are measurable. So $g(x) = \sup_{k \in \mathbb{N}} f_k(x)$ is measurable. Same for inf.

$$\{g > a\} = \bigcup_{k \in \mathbb{N}} \{f_k > a\}$$

$$\{g \geq a\}$$

$$\{g < a\}$$

$$\{g \leq a\}$$

First case done. Other follow similarly. HAND WAVE! For inf

$$\{\inf_k f_k < a\} = \bigcup_{k \in \mathbb{N}} \{f_k < a\}.$$

\square

Theorem 2.3. *Suppose $f_k : E \rightarrow \overline{\mathbb{R}}$ are measurable, then are measurable.*

Proof. Follows immediately from preceding result. [$\lim = \limsup = \liminf$ when it exists.] \square

Approximation: Given measurable f , find ‘simple’ functions f_k such that $f_k \rightarrow f$ pointwise. Here by simple we mean finite range.

Definition (Simple). g is simple if its range is finite subset of \mathbb{R} . [Need measurable?]

Need this to build notion of integration.

Theorem 2.4. For all measurable functions, there exists f_k such that $f_k \rightarrow f$ pointwise. If $f \geq 0$, we can take $f_1 \leq f_2 \leq \dots$.

Proof. $2^{-k} \text{floor}(2^k f)$. (floor to $\frac{\text{integer}}{2^k}$). This converges monotonically since increasing k shortens the rounding distance, and then not as much difference so this converges to f uniformly. Range not finite but countable. So just need to distribute the values among many functions.

$$f_k(x) = \begin{cases} \frac{j-1}{2^k}, & \frac{j-1}{2^k} \leq f(x) < \frac{j}{2^k} \\ k, & f(x) \geq k \end{cases}$$

f_k simple and $f_k \rightarrow f$. Now $f_k = \min(k, \max(-k, 2^{-k} \text{floor}(2^k f)))$. $f \geq 0$, then $f_1 \leq f_2 \leq \dots$. \square

Definition. $f : E \rightarrow \mathbb{R}$ is continuous at $a \in E$ if for all $\epsilon > 0$, there exists $\delta > 0$ such that if $x \in E$ and $|x - a| < \delta$, then $|f(x) - f(a)| < \epsilon$.

Why better than normal definition with limit? Because a can be a limit point not defined there. Take $f(x) = 1/x^2$, rather like allowing $f(0) = \infty$.

Semincontinuity means rather than bounding difference from above below, means one of inequality $-\epsilon < f(x) - f(a) < \epsilon$ holds.

Definition (Upper Semicontinuous (usc)). $f : E \rightarrow \overline{\mathbb{R}}$ is upper semicontinuous at $a \in E$ if for all $y > f(a)$, there exists $\delta > 0$ such that if $|x - a| < \delta$, then $f(x) < y$.

Definition (Lower Semicontinuous (lsc)). $f : E \rightarrow \overline{\mathbb{R}}$ is lower semicontinuous at $a \in E$ if for all $y < f(a)$, there exists $\delta > 0$ such that if $|x - a| < \delta$, then $f(x) > y$.

Example 2.3.

$$f(x) = \begin{cases} \frac{1}{x^2}, & x \neq 0 \\ \infty, & x = 0 \end{cases}$$

This is usc (automatically because of the ∞ value, why?) and lsc.

Both types of these functions are measurable.

Remark: lsc \rightarrow measurable usc \rightarrow measurable

Proof. if f usc on measurable E , then $\{f > a\}$ is open in E . So $\{f > a\} = E \cap U$. Hence $\{f > a\}$ is measurable.

For USC, use $\{f < a\}$.

BV function

Given $f : [a, b] \rightarrow \mathbb{R}$, define $V(f; a, b) = \sup_p \sum_{i=1}^n |f(x_i) - f(x_{i-1})|$ sup over all partitions $P : a = x_0 < x_1 < \dots < x_n = b$. $V(x^2; -1, 1) = 2$ ($P = \{-1, 0, 1\}$ gives the maximum value.)

Definition. f is BV on $[a, b]$ if $V(f; a, b) < \infty$

Interesting not called finite variation.

Something which is not BV is

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

Theorem 2.5. If f is continuous on $[a, b]$ and differentiable on (a, b) and $\int_a^b |f'| dx$ (riemann integral), then $V(f; a, b) = \int_a^b |f'| dx$.

C.f. examples above.

Proof. Mean Value Theorem.

$$\sum_i |f(x_i) - f(x_{i-1})| = \sum_i |f'(\xi_i)|(x_i - x_{i-1})$$

$x_{i-1} < \xi_i < x_i$. the above converges to integral. □

3. Subadditive wrt f : $V(f + g; a, b) \leq V(f; a, b) + V(g; a, b)$.

Proof. Triangle inequality.

$$|f(x_i) + g(x_i) - (f(x_{i-1}) + g(x_{i-1}))|$$

then use triangle splits into two and done.

Why not exactly like integral, i.e. strict inequality. Take nonconstant function f and then $g = -f$. Note $f + g$ identically zero.

Sum of monotone functions are bounded variation.

Theorem 2.6 (BV Decomposition). If f is BV on $[a, b]$, then there exists increasing g, h on $[a, b]$ such that $f = g - h$.

Proof. Let $g(x) = V(f; a, x)$, increasing (monotonicity wrt interval). This forces $h(x) := g(x) - f(x)$. Why is $h(x) \leq h(y)$ for $x \leq y$? $V(f; a, x) - f(x) \leq V(f; a, y) - f(y)$. Why? $f(y) - f(x) \leq V(f; a, y) - V(f; a, x) = V(f; x, y)$, clearly this right side is $\geq |f(x) - f(y)|$. □

Theorem 2.7 (Egorov's Theorem). Suppose $\{f_k\}$ is a collection of measurable functions, $f_k : E \rightarrow \mathbb{R}$, $|E| < \infty$, and $f_k \rightarrow f$ a.e., where $f : E \rightarrow \mathbb{R}$. Then for all $\epsilon > 0$, there exists a closed set $E(\epsilon) \subset E$ such that $f_k \rightarrow f$ uniformly on $E(\epsilon)$ and $|E \setminus E(\epsilon)| < \epsilon$.

Proof. We need to find a set on which $\{f_k\}$ converges uniformly. It suffices to find, for each $j \in \mathbb{N}$, $E_j \subset E$ such that $|E \setminus E_j| < \epsilon/2^j$, and $\sup_{E_j} |f - f_k| \leq 1/j$ for sufficiently large k . Indeed, letting $E(\epsilon) := \bigcap_j E_j$, then $f_k \rightarrow f$ uniformly on $E(\epsilon)$, and $|E \setminus E(\epsilon)| \leq \sum_j \epsilon/2^j < \epsilon$.

To find E_j , consider the set $G_m := \{x : |f_k(x) - f(x)| < 1/j \text{ for all } k \geq m\}$. Since $|f_k - f| \rightarrow 0$ a.e., $\bigcup_m G_m$ is almost E , i.e. $|\bigcup_m G_m| = |E|$. By continuity of measure for nested unions, $|G_m| \rightarrow |E|$. Therefore, there exists m such that $|E \setminus G_m| < \epsilon/2^j$. Defining $E_j := G_m$ completes the proof. \square

Theorem 2.8 (Lusin's Theorem). Let E be a measurable set. A function $f : E \rightarrow \mathbb{R}$ is measurable if and only if for all $\epsilon > 0$, there exists a closed set $E(\epsilon) \subset E$ such that $f|_{E(\epsilon)}$ is continuous, and $|E \setminus E(\epsilon)| < \epsilon$.

Proof. Suppose that f is a simple function, i.e. $f = \sum_{k=1}^n a_k \chi_{E_k}$, where E_1, \dots, E_n are disjoint measurable sets. Choose a closed subset $E'_k \subset E_k$ such that $|E_k \setminus E'_k| < \epsilon/n$. Then $E(\epsilon) := \bigcup_{k=1}^n E'_k$ so that $|E \setminus E(\epsilon)| < \epsilon$. It only remains to check that $f|_{E(\epsilon)}$ is continuous, but this is immediate as the preimage of any set is closed.

Now suppose only that $|E| < \infty$. There is a collection of simple functions $\{f_k\}$ such that $f_k \rightarrow f$ pointwise. Choose some $E(\epsilon/2^k)$ for f_k as in the case above. By Egorov's Theorem, $f_k \rightarrow f$ uniformly on a set $E(\epsilon)$. Now f is continuous on $E(\epsilon) \cap (\bigcap_k E_k(\epsilon/2^k))$, which has complement of measure at most $\epsilon + \sum_k \epsilon/2^k = 2\epsilon$.

Now suppose that $|E| = \infty$. Write $E = \bigcup [E \cap \{|j| \leq |x| \leq j+1\}]$. In each shell, we have $E_j(\epsilon/2^j)$, closed set such that $f|_{E_j(\epsilon/2^j)}$ is continuous. Note that f continuous on $\bigcup_j E_j(\epsilon/2^j)$. \square

The converse direction of Lusin's theorem is simple since $f|_{E(1/j)}$ is continuous, then we have f Borel on $\bigcup E(1/j)$ simply because $\{x \in \bigcup E(1/j) : f > a\} = \bigcup \{x \in E(1/j) : f > a\}$, countable union of Borel sets. and the rest has measure 0.

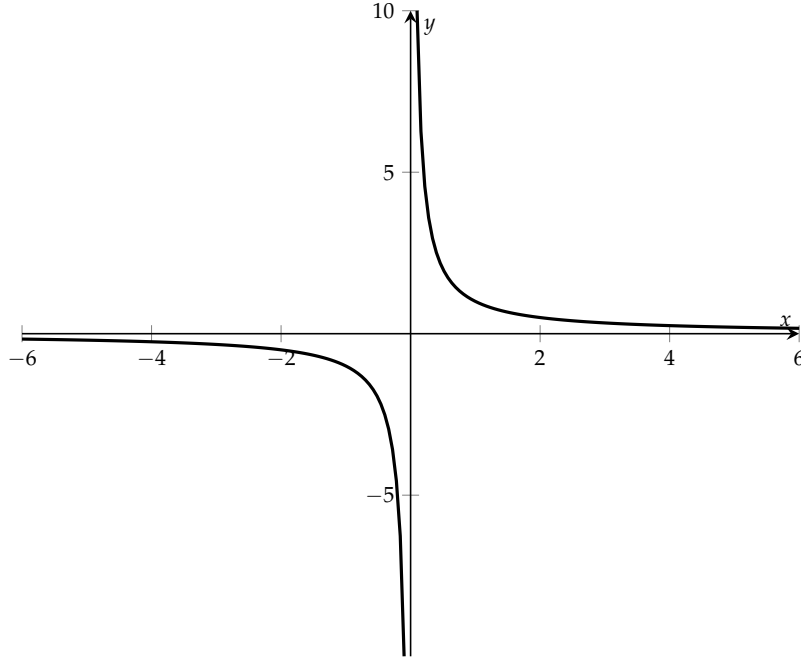
2.3 Convergence in Measure

Recall $\limsup E_k := \bigcap_{m=1}^{\infty} \bigcup_{k=m}^{\infty} E_k = \{x : x \in E_k \text{ infinitely many } k\}$. We will not need \liminf quite as regularly as \limsup .

Definition (Convergence in Measure). Let $\{f_k\}$ be a collection of measurable functions, $f_k : E \rightarrow \mathbb{R}$. We say f_k converges to f in measure if for all $\epsilon > 0$, $|\{f - f_k| > \epsilon\}| \rightarrow 0$. We denote this by $f_k \xrightarrow{m} f$.

If we expand this definition out more, we have the following: $f_k \xrightarrow{m} f$ if for $\epsilon > 0$ and for all $\epsilon_2 > 0$, there exists N such that $|\{ |f - f_k| > \epsilon \}| < \epsilon_2$ for all $k \geq N$. If this is possible, one may as well replace the double epsilon with a single epsilon for a more concise version. So the definition could be taken as follows: $f_k \xrightarrow{m} f$ if for all $\epsilon > 0$, there exists N such that $|\{ |f_k - f| > \epsilon \}| < \epsilon$ for all $k \geq N$.

Example 2.4. $f_k(x) = \frac{1}{kx} \xrightarrow{m} 0$ on \mathbb{R} .



We know that $\{ |f_k| > \epsilon \} = \{ |x| < \frac{1}{k\epsilon} \}$. This set has measure $2/(k\epsilon)$, which tends to zero as k tends to infinity. \triangleleft

Example 2.5. $f_k(x) = \frac{x}{k} \not\xrightarrow{m} 0$ on \mathbb{R} . $\{ |f_k| > \epsilon \} = \{ \frac{|x|}{k} > \epsilon \} = \{ |x| > k\epsilon \}$, which has infinite measure. \triangleleft

Example 2.6. Convergence in measure does not imply pointwise convergence. Let $f_k = \chi_{I_k}$, where I_k are intervals.

$[0,1]$ measure 1 $[0,1/2]$, $[1/2,1]$ measure 1/2 $[0,1/3]$, $[1/3,2/3]$, $[2/3,1]$ measure 1/3

For all $\epsilon > 0$, $\{ |f_k| > \epsilon \} < |I_k| \rightarrow 0$.

So the sequence converges to 0 in measure. However, it clearly does not converge to 0 in measure.

There are infinitely many times point part of the interval. That is for all $x \in (0,1)$, there are infinitely many k such that $x \in I_k$ so that $f_k(x) \not\rightarrow 0$. This can be expressed as saying $\limsup I_k = [0,1]$.

Theorem 2.9. If $|E| < \infty$, then convergence a.e. implies convergence in measure.

Proof. Given $\epsilon > 0$, Egorov's Theorem gives $E(\epsilon)$ such that $|E \setminus E(\epsilon)| < \epsilon$, and $f_k \rightarrow f$ uniformly on $E(\epsilon)$. Then there exists N such that $|f_k - f| < \epsilon$ on $E(\epsilon)$ for all $k \geq N$. But then $|\{|f_k - f| > \epsilon\}| < \epsilon$, which implies convergence in measure. \square

Theorem 2.10. If $f_k \xrightarrow{m} f$, then there is a subsequence $\{f_{k_j}\}$ converging to f a.e..

Proof. Choose $\epsilon = 1/2^j$. Find N_j such that $|\{|f_k - f| > 1/2^j\}| < 1/2^j$ for all $k \geq N$ (can make $N_{j+1} > N_j$). We have $\{f_{N_j}\}$. We claim that $f_{N_j} \rightarrow f$ a.e.. Let $E_j = \{|f_{N_j} - f| > 1/2^j\}$. Since $|E_j| < 1/2^j$, we know $\sum |E_j| < \infty$. But then by HW 1, we know $|\limsup E_j| = 0$. For all $x \notin \limsup E_j$, we have $x \in E_j$ only finitely many times. Then for sufficiently large j , $x \notin E_j$, say $x \notin E_j$ for $j \geq j_0$. But then $|f_{N_j}(x) - f(x)| < 1/2^j \rightarrow 0$. \square

Remark. "Triangle Inequality" (quasi). Convergence in measure is not like convergence in metric space, though it is tempting to think of it in this way. We do have a replacement for a parallel triangle inequality. If $|\{|f - g| > \epsilon/2\}| < \epsilon/2$, and $|\{|g - h| > \epsilon/2\}| < \epsilon/2$, then $|\{|f - h| > \epsilon\}| < \epsilon$. The proof is routine: $|f - h| \leq |f - g| + |g - h|$, so $|f - h| > \epsilon$ then one of the right side must be at least $\epsilon/2$. Then $\{|f - h| > \epsilon\} \subset \{|f - g| > \epsilon/2\} \cup \{|g - h| > \epsilon/2\}$. Their measures being at most $\epsilon/2$, the result then follows.

Definition (Cauchy in Measure). A collection of measurable functions $\{f_k\}$ is said to be Cauchy in measure if for all $\epsilon > 0$, there exists N such that $|\{|f_j - f_k| > \epsilon\}| < \epsilon$ for all $k, j \geq N$.

As it turns out, this is equivalent to convergence in measure.

Theorem 2.11. Cauchy in measure is equivalent to convergence in measure.

Proof (Sketch). \leftarrow is immediate by "triangle inequality" measure remark. Now the other direction. There are three standard steps that apply in these types of arguments. The strategy being the important piece here. Step 1. Enough to obtain a convergent subsequence: f_1, f_2, f_3, \dots cauchy sequence of functions with convergent subsequence. Use triangle inequality to show must converge to limit of conv subsequence. Conclude $f_k \rightarrow f$. Notice this holds for any type of convergence in any space. Step 2: Every Cauchy sequence has a "geometric" subsequence. Meaning $|\{|f_{k_j} - f_{k_{j+1}}| > 1/2^j\}| < 1/2^j$. This is true for any metric space and any convergence. Use $\epsilon = 1/2^j$, and get k_j . Step 3. Geometric trick implies convergent: given $\epsilon > 0$, let m be such that $1/2^{m-1} < \epsilon$. Let $A = \bigcup_{j \geq m} \{|f_{k_j} - f_{k_{j+1}}| > 1/2^j\}$. Note $|A| < \sum_{j=m}^{\infty} 1/2^j = 1/2^{m-1} < \epsilon$. On $E \setminus A$, we have $|f_{k_j} - f_{k_{j+1}}| \leq 1/2^j$ for all $j \geq m$. But this implies uniform convergence on the set $E \setminus A$. But then difference everywhere at most ϵ on $E \setminus A$ for sufficiently large j . Therefore, $|\{|f_{k_j} - f| > \epsilon\}| \leq |A| < \epsilon$. \square

3 Lebesgue Integral

Two approaches to $\int_E f, f \geq 0$ on E .

First, $\int_E f = \sup\{\int_E g : g \text{ simple } g \leq f\}, g = \sum_{k=1}^m a_k \chi_{E_k}$, where E_k are disjoint, measurable sets. Here $\int_E g = \sum a_k |E_k|$.

Second, $\int_E f = |R(f, E)|$, where $R(f, E) = \{(x, y) \in \mathbb{R}^{n+1} : x \in E, 0 \leq y \leq f(x)\}$.

First approach is common in many texts, we shall take the second.

So $\int_E f$ exists if and only if $R(f, E)$ is measurable.

Some advantages of the second approach are monotonicity and countable additivity over E : so $0 \leq f \leq g$ and $\int_E f, \int_E g$ exists, then $\int_E f \leq \int_E g$ because $R(f, E) \subset R(g, E)$.

Also if $E = \bigcup_{k \in \mathbb{N}} E_k$, disjoint measurable sets, and $\int_{E_k} f$ exists, then $\int_E f$ exists and $\int_E f = \sum_{k \in \mathbb{N}} \int_{E_k} f$ because $R(f, E) = \bigcup R(f, E_k)$, the union taken over the disjoint, measurable sets. Can show in other definition but takes work due to sup, whereas here we get it immediately. But we have the issue of existence and measurability.

The following main theorem.

Theorem 3.1. *If $f|_E \rightarrow [0, \infty]$ is measurable, then $\int_E f$ exists (possibility ∞).*

Idea of proof is from 4.1, $f = \lim f_k, f_1 \leq f_2 \leq \dots$ are simple. So we will show $R(f, E)$ are measurable and take their union. Need Lemma 1, $R(f_k, E)$ is measurable. Lemma 2 $\Gamma(f, E)$, which needs to be defined (simply the graph $\{(x, f(x)) : x \in E\}$) has measure 0. This is boundary top of graph of shaded area of graph, we care because $\bigcup R(f_k, E)$ need not generally cover $R(f, E)$. It will surely cover $R(f, E) \setminus \Gamma(f, E)$, then $y < f(x)$, so there exists k such that $y < f_k(x)$ so that $(x, y) \in R(f_k, E)$.

Both lemmas rely on

Lemma 3.1. *If $E \subseteq \mathbb{R}^n$ is measurable, then $E_a := E \times [a, b]$ is measurable in \mathbb{R}^{n+1} with measure $|E_a| = |E| \cdot a$, noting that $0 \cdot \infty = 0$ in measure theory.*

Proof. If E is an interval, follows by volume formula. If E is general, then tile it by intervals, half open (so intervals disjoint).

If E is G_δ of finite measure, write $E = \bigcap G_k, G_1 \supset \dots$ are open.

$$E_a = \bigcap_k (G_k \times [0, a]) \text{ so } |E_a| = \lim |G_k \times [0, a]| = a \lim |G_k| = a \cdot |E|.$$

What about general set? $|E| < \infty, E = H \setminus Z, M$ is $G_\delta, |M| = |E|, |H_a| = |H| \cdot a = |E| \cdot a$. So $|E_a| \leq |E| \cdot a$.

Also for all $\epsilon > 0$, there exists closed $F \subset E$ (hence G_δ), $|F| > |E| - \epsilon$.

$$|F_a| = a|F| > a|E| - a\epsilon$$

So E_a contains F_a closed of measure $> |E_a| - a\epsilon$. Hence E_a is measurable. $|E_a| = a|E|$.

Still assumes $a < \infty$, otherwise $a = k$ and let $k \rightarrow \infty$

□?

Lemma 3.2. $R(f_k, E) = \bigcup_j (E_j \times [0, a_j])$, E_j disjoint measurable sets, where $f_k = \sum_j a_j \chi_{E_j}$.

Note that this shows that if f takes countably many values, $f = a_k$ on E_k , then $\int_E f = \sum a_k |E_k|$.

Lemma 3.3. $|\Gamma(f, E)| = 0$. Let $g = 2^{-j} \text{floor}(2^j f)$, then $g \leq f \leq g + 2^{-j}$. Then $R(g, E) \subset R(f, E) \subset \text{????}$. Then $\Gamma(f, E) \subset R(g + 2^{-j}, E) \setminus R(g - 2^{-j}, E)$. This right side has measure at most $2 \cdot 2^{-j} |E| \rightarrow 0$, hence $|\Gamma(f, E)| = 0$.

Last time: $\int_E f = |R(f, E)|$, $R(f, E) = \{(x, y) \in \mathbb{R}^{n+1} : 0 \leq y \leq f(x), x \in E\}$. $\Gamma(f, E) = \{(x, f(x)) \in \mathbb{R}^{n+1} : x \in E\}$.

$\int_E f = \sup\{\int g : g \leq f, g \text{ simple}\}$, $g = \sum a_k \chi_{E_k} \rightarrow \sum g = \sum a_k |E_k|$, then $R(f, E) = (E \cap F) \times [0, 1]$.

Recall $R(f, E)$ is approximated by $R(f_k, E)$:

$\bigcup_k R(f_k, E) \subset R(f, E) \subset [\bigcup_k R(f_k, E)] \cup \Gamma(f, E)$.

Lemma: $|\Gamma(f, E)| = 0$ for all measurable f .

Proof. Write $E = \bigcup E_m$, $|E_m| < \infty$

Suffices $|\Gamma(f, E_m)| = 0$ for all $|E_m| < \infty$. Let $\mathcal{A}_k = \{x \in E_m : \epsilon k \leq f(x) < \epsilon(k+1)\}$, $k = 0, 1, 2, \dots$. Then $\Gamma(f, E_m) \subset \bigcup_k \mathcal{A}_k + [\epsilon k, \epsilon(k+1)]$, then $|\Gamma(f, E_m)| \leq \sum_k |\mathcal{A}_k| \epsilon = \epsilon |E_m|$, then $|\Gamma(f, E_m)| = 0$.

Theorem 3.2 (Monotone Convergence Theorem). If $f_k \geq 0$ are measurable on E and $f_1 \leq f_2 \leq \dots$, then $\int_E \lim f_k = \lim \int_E f_k$.

Proof. $f = \lim f_k$, then $R(f, E) = [\bigcup R(f_k, E)] \cup \Gamma(f, E)$ and $\bigcup R(f_k, E)$ are nested. So $|R(f, E)| = \lim |R(f_k, E)|$ by continuity of measure. \square

Example 3.1. $\mathbb{Q} = \{q_1, q_2, \dots\}$. Let $f_k = \chi_{\{q_1, \dots, q_k\}}$. Then $f_k \nearrow f = \chi_{\mathbb{Q}}$. Monotone convergence theorem fails for Riemann integral, so $\int_{\mathbb{Q}} = \lim \int f_k = 0$.

Theorem 3.3. The two definitions of $\int_E f$ agree:

$$\int_E f = \sup\{\sum a_k |E_k| : g \leq f, g = \sum a_k \chi_{E_k}\}$$

Proof. Note $\sum a_k |E_k| = \int_E g$ by measure of product. Since $f \geq g$, we get $\int f \geq \int g$ so \geq holds. Reverse: pick $g_k \nearrow f$ pointwise. Then $\int g_k \rightarrow \int f$ so $\sup \int g_k \geq \int f$. \square

Example 3.2. $f_1 \geq f_2 \geq \dots$. Counterexample $f_k = \chi_{[k, \infty)}$. $\lim \int f_k = \infty$ but $\int \lim f_k = 0$. Similarly $\inf\{\int g : g \geq f, g \text{ simple}\} \neq \int f$ in general. if $g \geq f$ then $g > 0$ but being simple $g \geq c > 0$ then $\int g = \infty$.

Chebyshev's Inequality: Suppose f is nonnegative measurable on E . For all $\alpha > 0$
 $|\{x \in E: f(x) > \alpha\}| \leq \frac{1}{\alpha} \int_E f$.

Proof. Let $A = \{f > \alpha\}$. Then $f \geq \alpha \chi_A \rightarrow \int f \geq \int \alpha \chi_A = \alpha |A| \rightarrow |A| \leq \frac{1}{\alpha} \int f$.

Example 3.3. $f(x) = \frac{1}{\sqrt{x}}$ on $[0, 1]$. Let $c = \int_{(0,1)} f \rightarrow$ Chebyshev $|\{f > \alpha\}| \leq \frac{c}{\alpha}$. In fact,
 $\{f > \alpha\} = (0, 1/\alpha^2) \rightarrow$ measure is $1/\alpha^2$, if $\alpha \geq 1$.

Additivity over functions: $\int_E (f + g) = \int_E f + \int_E g$, $f, g \geq 0$ measurable.

Proof. Use simple functions, $f_k \nearrow f$, $g_k \nearrow g$. So it suffices to prove that $\int (f_k + g_k) = \int f_k + \int g_k$. $f_k = \sum a_j \chi_{E_j}$, $g_k = \sum b_j \chi_{F_j}$. Let $H = E_i \cap F_j$ note $\cup E_i = \cup F_i = E$. Then $f_k = \sum_{i,j} a_i \chi_{H_{ij}}$, $g_k = \sum b_j \chi_{H_{ij}}$. Then $f_k + g_k = \sum (a_i + b_j) \chi_{H_{ij}}$ then $\int (f_k + g_k) = \sum (a_i + b_j) |H_{ij}| = \sum a_i |H_{ij}| + \sum b_j |H_{ij}|$.

Countable subadditivity:

$\int \sum_{k=1}^{\infty} f_k = \sum_k \int_E f_k$, $f_k \geq 0$ measurable.

True for $\sum_{k=1}^N$ then $N \rightarrow \infty$ using MCT

Example 3.4. Let $\mathbb{Q} \cap [0, 1] = \{q_1, \dots\}$

$$f_k = \frac{1}{\sqrt{|x - q_k|}} \int_{[0,1]} f \leq c < \infty$$

$$f = \sum_{k=1}^{\infty} \frac{1}{k^2} \frac{1}{\sqrt{|x - q_k|}} \xrightarrow{\lim_{x \rightarrow q} f(x) = \infty \text{ for all } q \in \mathbb{Q}} \int f \leq c \sum \frac{1}{k^2} < \infty \text{ so } |\{f > \alpha\}| \leq \frac{1}{\alpha} C',$$

what does this say about \mathbb{Q} ?

We have not yet proved $\int_E cf = c \int_E f$, where $c \geq 0$ and $f \geq 0$.

Proposition 3.1.

$$\int_E cf = c \int_E f$$

Proof. This is true for simple functions: $\int c(\sum a_k \chi_{E_k}) = \int \sum ca_k \chi_{E_k} = \sum ca_k |E_k| = c \sum a_k |E_k| = c \int \sum a_k \chi_{E_k}$. Then for general functions use the Monotone Convergence Theorem. For general $f \geq 0$, take simple $f_k \geq 0$, $f_k \nearrow f$, then $cf_k \nearrow cf$, by MCT $\int cf = \lim cf_k = c \lim \int f_k = c \int f$. \square

What if the convergence is not $f_k \nearrow f$?

Example 3.5. $\chi_{[k, \infty)} \rightarrow 0$ pointwise. But $\int \chi_{[k, \infty)}$ are all infinite and certainly do not converge to 0. Note that $\chi_{[k, \infty)} \searrow 0$.

Example 3.6. $k \chi_{(0, 1/k)}$. But $\int k \chi_{(0, 1/k)} = 1 \not\rightarrow 0$.

Example 3.7. $\frac{1}{k} \chi_{[0, k]} \rightarrow 0$ uniformly but $\int \frac{1}{k} \chi_{[0, k]} = 1 \rightarrow 0$.

What is the similarity of these three examples? The functions vanish but their integrals do not???? The general principle is that the area under the graph can disappear or 'escape' to infinity but does not appear from nowhere..... A precise statement is Fatou's Lemma. As a reminder,

Definition (limsup, liminf). Given a sequence $\{f_k\}$,

$$\limsup_{k \rightarrow \infty} f_k = \lim_{n \rightarrow \infty} \sup_{k \geq n} f_k \quad \liminf_{k \rightarrow \infty} f_k = \lim_{n \rightarrow \infty} \inf_{k \geq n} f_k.$$

Example 3.8. (i) $\lim f_k(x)$ exists if and only if $\limsup f_k(x) = \liminf f_k(x)$.

(ii) The Piano Key Sequence (press one key going left to right): $\chi_{[0,1]}$, then $\chi_{[0,1/2]}, \chi_{[1/2,1]}$, then $\chi_{[0,1/3]}, \chi_{[1/3,2/3]}, \chi_{[2/3,1]}$. [Go in pascal triangle order]

$$\limsup f_k = \chi_{[0,1]}$$

$$\liminf f_k = 0$$

Remark. Relation to sequences to sets $\{E_k\}$: $\limsup E_k = \bigcap_{m=1}^{\infty} \bigcup_{k \geq m} E_k$, $\liminf E_k = \bigcup_{m=1}^{\infty} \bigcap_{k \geq m} E_k$. $\limsup \chi_{E_k} = \chi_{\limsup E_k}$ and same for liminf. and sup like union and inf like intersection.

Lemma 3.4 (Fatou). $f_k \geq 0$ measurable.

$$\int_E \liminf f_k \leq \liminf \int_E f_k$$

Proof. Let $g_m = \inf_{k \geq m} f_k$. Then $g_m \nearrow \liminf f_k$. By MCT, $\int g_m \rightarrow \int \liminf f_k$. But $f_k \geq g_m$ so that $\int f_k \geq \int g_m$, hence $\liminf \int f_k \geq \lim \int g_m$. [Rem: if $a_k \geq b_k$ for all k , then $\liminf a_k \geq \liminf b_k$, $\limsup a_k \geq \limsup b_k$.] \square

Theorem 3.4 (Lebesgue Dominated Convergence (DCT)). Suppose $f_k \geq 0$ are measurable and there is $\phi \geq 0$ measurable such that $f_k \leq \phi$ (ϕ dominates) and $\int_E \phi < \infty$. If f_k converges to f a.e., then $\int_E f_k \rightarrow \int_E f$.

Proof. Fatou's Lemma gives $\int f \leq \liminf \int f_k$. Also, $\phi - f_k \geq 0$. Then Fatou's Lemma applies again: $\int \phi - f \leq \liminf \int (\phi - f_k)$. Which is $\int f \geq \limsup \int f_k$. Then $\int \phi - \int f \leq \liminf (\int \phi - \int f_k)$ cancel constant $-\int f \leq \liminf (-\int f_k)$ then $\int f \geq \limsup \int f_k$. Comparing very first and last equations, we have $\int f = \lim_{k \rightarrow \infty} \int f_k$. \square

Example 3.9. $\frac{1}{k} \chi_{[0,k]} \rightarrow 0$ uniformly.

Try to cover with one function get something like $1/x$. This is not good enough for DCT since $\int \frac{1}{x} = \infty$. If $f_k \leq \frac{1}{x^2}$ on $[1, \infty)$, this examples does not work. 'Best' you can do is $\frac{1}{k^2} \chi_{[0,k]} \rightarrow 0 = \int \lim f_k$.

Example 3.10. $\lim \int_{[1,\infty)} \frac{\sin^{2k} x}{x^2} dx$. Now $\frac{\sin^{2k} x}{x^2} \rightarrow 0$ a.e. and is at most $\frac{1}{x^2}$ in absolute value, so must be 0. Now if $\lim \int_{[0,\infty)} \frac{\sin^{2k} x}{x^2} dx$, $\frac{\sin^{2k} x}{x^2} \leq \frac{\sin^2 x}{x^2}$ and $\int_{[0,1]} \frac{\sin^2 x}{x^2} \leq \int_{[0,1]} 1 \leq 1$. so $\int_{[0,\infty)} \frac{\sin^2 x}{x^2} < \infty$.

Suppose that $f : E \rightarrow \overline{\mathbb{R}}$ is measurable. Define functions $f^+ = \max(f, 0)$ and $f^- = \max(-f, 0)$. These are both nonnegative functions with $f^+ - f^- = f$ and $f^+ + f^- = |f|$.

Definition (Integrable Function). Suppose $f : E \rightarrow \overline{\mathbb{R}}$ is a measurable function, not necessarily nonnegative. We define $\int_E f$ to be

$$\int_E f := \int_E f^+ - \int_E f^-,$$

provided at least one of the integrals on the right is finite. We say that f is integrable on E if $\int_E f$ is finite, and we write $f \in L^1(E)$,

Remark. Note that $\int f$ is finite if and only if $\int f^+$ and $\int f^-$ are finite if and only if $\int |f|$ is finite. Therefore, all Lebesgue integrable functions with finite integral are automatically absolutely integrable. This means there are integrals which are (Improper) Riemann integrable which are not Lebesgue integrable. Take Example ??, $\int_{(0,\infty)} \frac{\sin x}{x} dx$ does not exist as a Lebesgue integrable, i.e. $\frac{\sin x}{x} \notin L^1(0, \infty)$, while it is Riemann integrable on $(0, \infty)$.

We next examine a few properties of this integral. Not surprisingly, many of these properties resemble those of their primordial Riemann integral.

- Triangle Inequality: $|\int_E f| \leq \int_E |f|$ as $|\int f^+ - \int f^-| \leq \int f^+ + \int f^-$.
- Additivity of Domain: If $\{E_k\}_{k=1}^N$ is a collection of disjoint, measurable sets, then $\int_{\cup E_k} f = \sum_{k=1}^N \int_{E_k} f$ as $\int f^+$, $\int f^-$ split using the properties of the integral for nonnegative functions.

However, care is needed even in the countable additivity case. Consider $\int_0^\infty \sin x$, defining the E_k to be as below:

We know that $\int_{E_k} f = 0$ for all k but that $\int_{\cup E_k} f$ does not exist.

Lemma 3.5. Suppose that $\int_E f$ exists. Let $E = \cup_{k=1}^\infty E_k$ be a union of disjoint, measurable sets. Then $\int_{E_k} f$ exists for all k , and

$$\int_E f = \sum_{k=1}^\infty \int_{E_k} f$$

Proof. By countable additivity for integrals of nonnegative functions, we know that $\int_E f^+ = \sum_k \int_{E_k} f^+$ and $\int_E f^- = \sum_k \int_{E_k} f^-$. Now observe

$$\int_E f = \sum_k \left(\int_{E_k} f^+ - \int_{E_k} f^- \right).$$

Since $\int_E f$ exists, at least one of $\int_E f^+$, $\int_E f^-$ is finite. □

Additivity over f :

Suppose we have $\int_E (f + g) = \int_E f + \int_E g$, assuming that both integrals on the right are finite. Let $h := f + g$. There are eight possibilities for the signs of (f, g, h) . This splits E into eight different parts. On each part, write a between f, g, h as addition of nonnegative functions, e.g. $f \geq 0, g < 0, h < 0$, then $f + g = h$ becomes $f + (-h) = (-g)$, all nonnegative here. Then

$$\int f + \int (-h) = \int (-g)$$

Note that we are making use of the fact that the functions are finite a.e..

Remark: if $f \in L^1(E)$, then $|f| < \infty$ a.e., since $\int |f| < \infty$.

Convergence:

There is little to say about convergence since everything follows from convergence of nonnegative functions.

For example, how does MCT apply? The answer is that it does not really apply. We need an integrable 'baseline', $\phi \in L^1(E)$, i.e. a function to play the role of the x -axis.

MCT for nonnegative functions:

If $\phi \leq f_1 \leq f_2 \leq \dots$ or $\phi \geq f_1 \geq f_2 \geq \dots$, then $\int f_k \rightarrow \int \lim f_k$.

Proof. Use MCT for $f_k - \phi$ or $\phi - f_k$, nonnegative, increasing. Then $\int (f_k - \phi) \rightarrow \int (f - \phi)$; hence, $\int f_k \rightarrow \int f$. □

However, there is one useful result here:

Theorem 3.5 (Dominated Convergence Theorem). *If $f_k \rightarrow f$ a.e. on E , and are measurable, and there exists $\phi \in L^1(E)$ such that $|f_k| \leq \phi$ for all k , then*

$$\int_E f_k \longrightarrow \int_E f.$$

Proof. Since $0 \leq \underbrace{f_k + \phi}_{\geq 0} \leq 2\phi$, DCT for nonnegative functions applies. $\int (f_k + \phi) \rightarrow \int (f + \phi)$. □

Example 3.11. Consider $\sum_{k=1}^{\infty} B_k \sin(kx)$ on the interval $(0, \pi)$. Assume that $C := \sum |B_k| < \infty$. What do we need to assume about B_k to obtain

$$\int_{(0,\pi)} f(x) = \sum_k \int_{(0,\pi)} B_k \sin(kx)?$$

Observe that C is a domination function. $f_k(x) = \sum_{j=1}^k B_j \sin(jx)$ satisfies $|f_k| \leq C$. Also, $f_k \rightarrow f$ pointwise.

Shorter proof: let $h = f + g$: $h^+ - h^- = f^+ - f^- + g^+ - g^-$. $h^+ + f^- + g^- = f^+ + g^+ + h^-$. Integrate, apply additivity for nonnegative. $\int h^+ + \int f^- + \int g^- = \int f^+ + \int g^+ + \int h^-$. All integrals finite then $\int h^+ - \int h^- = \int f^+ - \int f^- + \int g^+ - \int g^-$.

DCT: Works for continuous parameters.

If we have f_t (t real) and $|f_t| \leq \phi$, $\phi \in L^1(E)$ and f_t measurable and $f_t \rightarrow f$ a.e. as $t \rightarrow t_0$. Then $\lim_{t \rightarrow t_0} \int_E f_t = \int_E f$.

Proof. Sequential characterization of limits: Take any sequence $t_k \rightarrow t_0$ and apply DCT to f_{t_k} . Get $\int f_{t_k} \rightarrow \int f$. \square

Example 3.12. Suppose $|E| < \infty$, $f \in L^1(E)$, $f > 0$ on E . a) $\int_E f^p \rightarrow \int_E f$ as $p \rightarrow 0^+$ b) if also $\log f \in L^1(E)$, then $\int_E \frac{f^p - 1}{p} \rightarrow \int_E \log f$ as $p \rightarrow 0^+$.

apf) $f^p \rightarrow 1$ as $p \rightarrow 0^+$. Need dominating function. Consider only $0 < p \leq 1$. Then $f^p \leq f$ if $f \geq 1$ but notice $f^p \leq 1$ if $0 < f < 1$. So $f^p \leq \max(f, 1) \leq f + 1$. This is simpler than the less mysterious max. So $f + 1 \in L^1(E)$ dominating function. Hence, $\int_E f^p \rightarrow \int_E 1 = |E|$.

bpf) derivative of $p \mapsto \int_E f^p$ at $p = 0$ which is $\log f$. $\frac{f^p - 1}{p} \rightarrow \log f$. Slope of secant

Slope inc: $\frac{f^p - 1}{p} \leq \frac{f^1 - 1}{1} = f - 1$

$f \leq 1$: $\left| \frac{f^p - 1}{p} \right| \leq |\log f|$.

If $f < 0$ is measurable and $f \in L^1(E)$, then

$$\int_E \frac{f^p - 1}{p} \rightarrow \int_E \log f \text{ as } p \rightarrow 0^+.$$

We need integrability to make this possible.

MCT applies: if $\phi \in L^1$ is a 'baseline' and $\phi \geq f_1 \geq f_2 \geq \dots$ or $\phi \leq f_1 \leq f_2 \leq \dots$, then $\int \lim f_k = \lim \int f_k$.

For all $p_k \searrow 0$, we want to show $f_k = \frac{f^{p_k} - 1}{p_k}$ is a decreasing sequence, $f_1 \geq f_2 \geq f_3 \geq \dots$. The reason why is because of the shape of these things.

Now taking p_k which is smaller, so p_{k+1} is to left so the slope of the secant line is less because it is a convex function. The conclusion is the same for the decreasing version because this is also convex. 'Baseline' is then $f - 1$.

3.1 Relation between the Lebesgue and Riemann-Stieltjes Integrals

Recall most important HW (5.1#1): if $f \geq 0$ measurable, then $f \in L^1$ if and only if $\sum_{j=-\infty}^{\infty} 2^j |\{f > 2^j\}| < \infty$. Let $\omega_f = |\{x \in E: f(x) > \alpha\}|$. Notice this is just notation for something we have already discussed at length. With this notation, we can say that

$$f \in L^1(E) \iff \sum_{j=-\infty}^{\infty} 2^j \omega_{|f|}(2^j) < \infty,$$

where f is measurable but not necessarily nonnegative. What is the significance of the '2' here? The proof of the exercise only makes use of the fact that multiplication by 2 preserves the inequality: $g = 2^j \chi_{\{2^j < f < 2^{j+1}\}}$ and then $g \leq f \leq 2g$. This is like the Cauchy Condensation Test. However, this works for any real number $\lambda > 1$. This allows us to then characterize the above as

$$\begin{aligned} f \in L^1(E) &\iff \sum_{j=-\infty}^{\infty} 2^j \omega_{|f|}(2^j) < \infty \\ &\iff \sum_{j=-\infty}^{\infty} \lambda^j \omega_{|f|}(\lambda^j) < \infty \end{aligned}$$

for any $\lambda > 1$. The generalization of this is as follows:

Definition (L^p). $f : E \rightarrow \overline{\mathbb{R}}$ belongs to $L^p(E)$ if $\int |f|^p < \infty$, where $0 < p < \infty$.

By the discussion above (or #1 of HW 5.1), we know that $f \in L^p$ if and only if $\sum \lambda^j |\{|f|^p > \lambda^j\}| < \infty$, i.e. $\sum \lambda^j |\{f > \lambda^{j/p}\}| < \infty$. If we choose $\lambda = 2^p$, then

$$f^p \in L^1 \iff \sum_{j=-\infty}^{\infty} 2^{pj} |\{f > 2^j\}| < \infty.$$

Theorem 3.6. If $|E| < \infty$ and $f : (a, b]$ is measurable, where $a, b \in \mathbb{R}$, then

$$\int_E f = \int_a^b \alpha d(-\omega_f(\alpha)).$$

Proof. Consider \mathcal{P} : $a = x_0 < x_1 < \dots < x_n = b$. Let $E_i = \{x_{i-1} < f \leq x_i\}$. Note E_i is a partition of E .

$$\sum_i x_{i-1} \chi_{E_i} \leq f \leq \sum_i x_i \chi_{E_i}.$$

But then

$$\sum_{i=1}^n x_{i-1} |E_i| \leq \int_E f \leq \sum_{i=1}^n nx_i |E_i| (*)$$

Here

$$|E_i| = -(\underbrace{\omega(x_i)}_{f > x_i} - \underbrace{\omega(x_{i-1})}_{f > x_{i-1}})$$

But then the left sum is $L(\alpha, P, -\omega)$ and the right one is $U(\alpha, P, -\omega)$. The sup of L and inf of U both go to $\int_a^b \alpha d(-\omega)$. Thus, $\int_E f = \int_a^b f d(-\omega)$. \square

3.2 Relation between the Lebesgue and Riemann Integral

We want to compare $\int_{[a,b]} f$ and $R\int_a^b f$, where $a, b \in \mathbb{R}$ and f is bounded. Following shows we have generalized.

Theorem 3.7. *If $R\int_a^b f$ exists, then $\int_{[a,b]} f$ exists and they are equal.*

Proof. Riemann then Leb: If R integral exists, then there exists sequence of partitions $\{P_k\}$ such that $U(f, P_k)$ and $L(f, P_k)$ converge to $\int_a^b f$. Arrange $P_1 \subset P_2 \subset P_3 \subset \dots$. Let g_k, h_k be step functions for these partitions. $g_k \leq f \leq h_k$ and $L(f, P_k) = \int g_k$, $U(f, P_k) = \int h_k$.

By MCT, $g_1 \leq g_2 \leq \dots$ and $h_1 \geq h_2 \geq \dots$. Then $\int g_k \rightarrow \int g$, where $g = \lim g_k$ and $\int h_k \rightarrow \int h$, where $h = \lim h_k$. So $\int g = \int h = R\int_a^b f$ since U and L converge to $\int_a^b f$. But we know that $g \leq f \leq h$. This means that we must have equality a.e.. But this completes the proof. \square

The following characterization of Riemann integrals.

Theorem 3.8. *Riemann integral exists if and only if $|\mathcal{D}| = 0$, where \mathcal{D} is the set of discontinuities of f .*

Proof. See text.

4 Repeated Integration

4.1 Fubini's Theorem

Given the function $f(x, y) = \frac{1}{\sqrt{xy}}$ on the interval $[0, 1] \times [0, 1]$, we already have a concept of $\int_{[0,1] \times [0,1]} f$. But how does this integral relate to the integral

$$\int_0^1 \left(\int_0^1 \frac{1}{\sqrt{xy}} dy \right) dx?$$

One can ask the same questions and confront the same issues with infinite double summations, e.g. $\sum_{i,j} a_{ij}$.

Example 4.1. Consider the infinite matrix with 1s on the main diagonal and -1 on the superdiagonal and zeros elsewhere. The sum of each row is 0 and, except for the first column, the sum of column is 0. This implies

$$\begin{aligned} \sum_{i=1}^{\infty} \left(\sum_{j=1}^{\infty} a_{ij} \right) &= 0 \\ \sum_{j=1}^{\infty} \left(\sum_{i=1}^{\infty} a_{ij} \right) &= 1 \end{aligned}$$

One can create a parallel example with integrals. For example, take $f = 1, -1, 0$ and consider

$$\begin{aligned} \int_0^{\infty} \left(\int_0^{\infty} f(x, y) dx \right) dy &= 0 \\ \int_0^{\infty} \left(\int_0^{\infty} f(x, y) dy \right) dx &= 0 \end{aligned}$$

Note that $f \notin L^1([0, \infty) \times [0, \infty))$.

The goal of this section is to establish Fubini's Theorem, stated below, and its applications. This was proven by Guido Fubini in 1907. The theorem allows determines conditions under which it is possible to compute double integrals using iterated integrals. Consequently, this allows one to exchange the order of integration. In 1909, a student of Fubini named Leonida Tonelli established a similar theorem (Tonelli's Theorem) but applying to functions which are nonnegative instead of integrable.

Theorem 4.1 (Fubini's Theorem). *Let $I_1 \subseteq \mathbb{R}^n$, $I_2 \subseteq \mathbb{R}^m$ be intervals, and $I := I_1 \times I_2$. Suppose that $f \in L^1(I)$. Then*

- (i) for almost every $\mathbf{x} \in I_1$, the ‘slice’ function $\mathbf{y} \mapsto f(\mathbf{x}, \mathbf{y})$ is measurable and integrable on I_2 .
- (ii) the ‘once-integrated’ function $\mathbf{x} \mapsto \int_{I_2} f(\mathbf{x}, \mathbf{y}) \, d\mathbf{y}$ is measurable and integrable on I_1 , and its value is $\int_I f$, i.e.

$$\iint_I f(\mathbf{x}, \mathbf{y}) \, d\mathbf{x} d\mathbf{y} = \int_{I_1} \left(\int_{I_2} f(\mathbf{x}, \mathbf{y}) \, d\mathbf{y} \right) d\mathbf{x}.$$

The statement ‘for almost every \mathbf{x} ’ is required as a slice may have ‘issues’ on some set of measure 0, which will not affect the integral. As a parallel issue, for example, consider the matrix in Example 4.1. One could instead insert for some zero entry a nonmeasurable function or the value ∞ . So while the ‘slice’ function may be problematic, its integral is unchanged since its values were changed on a set of measure zero. Recall that

$$\int_I f = \int_{\mathbb{R}^{n+m}} f \cdot \chi_I.$$

By replacing f with $f\chi_I$, i.e. taking $f \equiv 0$ outside of I , we may reduce the proof of Fubini’s Theorem to the case where $I_1 = \mathbb{R}^n$, $I_2 = \mathbb{R}^m$, and $I = \mathbb{R}^{n+m}$. Let $\mathcal{F} := \{f \in L^1(\mathbb{R}^{n+m}) : \text{Fubini holds for } f\}$. To prove Fubini’s Theorem, we prove that $\mathcal{F} = L^1$. First, observe the following properties of \mathcal{F} :

Proposition 4.1. *Let $\mathcal{F} := \{f \in L^1(\mathbb{R}^{n+m}) : \text{Fubini holds for } f\}$. Then*

- (i) \mathcal{F} is a vector space: $f, g \in \mathcal{F}$, then $\alpha f + \beta g \in \mathcal{F}$.
- (ii) \mathcal{F} is closed under monotone limits: if $f_k \in \mathcal{F}$ and $f_k \nearrow f$ or $f_k \searrow f$, where $f \in L^1(\mathbb{R}^{n+m})$, then $f \in \mathcal{F}$.
- (iii) Squeeze Property: if $g \leq f \leq h$ and $g = h$ a.e., where $g, h \in \mathcal{F}$, then $f \in \mathcal{F}$.
- (iv) $\chi_I \in \mathcal{F}$ for any closed bounded interval I .

Proof.

- (i) This is immediate from the fact that if f, g are measurable, then $f + g$ and cf are measurable. Similarly, if $f, g \in L(E)$, then $f + g, cf \in L(E)$.
- (ii) By replacing f_k with $-f_k$ for all k , we may assume without loss of generality that $f_k \nearrow f$. For each k , there is a set $Z_k \subseteq \mathbb{R}^n$ with measure zero such that the \mathbf{x} -slice function of f_k is integrable for $\mathbf{x} \notin Z_k$. Let $Z = \cup_k Z_k$. Since Z is a countable union of sets with measure zero, Z has measure zero. If $\mathbf{x} \notin Z$, then the \mathbf{x} -slice function $f_k(\mathbf{x}, \mathbf{y})$ is integrable for all k , and by the MCT applied to $\{f_k(\mathbf{x}, \mathbf{y})\}$ as functions of \mathbf{y} , we have

$$h_k(\mathbf{x}) := \int f_k(\mathbf{x}, \mathbf{y}) \, d\mathbf{y} \nearrow \int f(\mathbf{x}, \mathbf{y}) \, d\mathbf{y} =: h(\mathbf{x}).$$

By assumption, $h_k(\mathbf{x})$ is integrable as a function of \mathbf{x} , $f_k \in L(I)$, and $\iint f_k(\mathbf{x}, \mathbf{y}) \, d\mathbf{x}d\mathbf{y} = \int h_k(\mathbf{x}) \, d\mathbf{x}$. But then by the MCT,

$$\iint f(\mathbf{x}, \mathbf{y}) \, d\mathbf{x}d\mathbf{y} = \int h(\mathbf{x}) \, d\mathbf{x}.$$

As $f \in L(I)$, we have $h \in L(I_1)$, implying that h is finite almost everywhere. Therefore, Fubini's Theorem holds for f also.

We now prove the second. For almost all x , the slice of g and h are both integrable, and since the slice of f is between, it is also integrable.

(Proof of 3???) Once integrated functions for g, f, h satisfy $\int g \, dy \leq \int f \, dy \leq \int h \, dy$ for almost all x . Integrate with respect to x

$$\left(\int g \, dy \right),$$

but then

$$\left(\int g \, dy \right) dx \leq \left(\int f \, dy \right) dx \leq \left(\int h \, dy \right) dx$$

But the left/right integrals are equal to $\int_I g$ and $\int_I h$, respectively. But these are equal since $g = h$ a.e.. Therefore, $\left(\int f \, dy \right) dx = \int_I f$.

Proof of 4: Slice function of χ_I is either $\equiv 0$ or χ_{I_2} , so $\int \chi_I = |I_2|\chi_{I_1}$. Then $\int \chi_I \, dy \, dx = |I_2||I_1| = |I| = \int \chi_I$. \square

We are now in a position to prove Fubini's Theorem.

Theorem 4.1 (Fubini's Theorem). *Let $I_1 \subseteq \mathbb{R}^n$, $I_2 \subseteq \mathbb{R}^m$ be intervals, and $I := I_1 \times I_2$. Suppose that $f \in L^1(I)$. Then*

- (i) *for almost every $\mathbf{x} \in I_1$, the 'slice' function $\mathbf{y} \mapsto f(\mathbf{x}, \mathbf{y})$ is measurable and integrable on I_2 .*
- (ii) *the 'once-integrated' function $\mathbf{x} \mapsto \int_{I_2} f(\mathbf{x}, \mathbf{y}) \, d\mathbf{y}$ is measurable and integrable on I_1 , and its value is $\int_I f$, i.e.*

$$\iint_I f(\mathbf{x}, \mathbf{y}) \, d\mathbf{x}d\mathbf{y} = \int_{I_1} \left(\int_{I_2} f(\mathbf{x}, \mathbf{y}) \, d\mathbf{y} \right) d\mathbf{x}.$$

Proof. We establish $\mathcal{F} = L^1$. We show this for G_δ sets and sets of measure 0, which will be enough.

a. First, we show that $X_I \in \mathcal{F}$ for any closed bounded interval. This does not necessary worked for closed. Indeed, open sets are union of closed sets.

Consider the sequence of characteristic functions on the boxes in the figure, say $\chi_{I_k} \nearrow \chi_I$. Now for a set which is not open or closed, squeeze this between an open set containing and a closed set inside it.

b. Now $\chi_G \in \mathcal{F}$ if G is open and $|G| < \infty$. Tile G by half-open cubes Q_j , $\chi_G = \sum \chi_{Q_j}$. Monotone limit of partials sums of 1 and 2 bullets above.

c. $\chi_H \in \mathcal{F}$ if H is a G_δ set with $|H| < \infty$. Indeed, there exists G open, $H \subset G$, and $|G| < |H| + \infty$ so if $H = \bigcap_{k=1}^\infty G_k$, G_k , with $H = (G \cap G_1) \cap (G \cap G_1 \cap G_2) \cap (G \cap G_1 \cap G_2 \cap G_3) \cap \dots$ so it is a nested intersection of open sets of finite measure. Then use second bullet above.

d. Now $\chi_Z \in \mathcal{F}$ if $|Z| = 0$. There exists a G_δ set H such that $Z \subset H$ and $|H| = 0$. Now $0 \leq \chi_Z \leq \chi_H$ and $\chi_H = 0$ a.e., then by squeeze (bullet 3) have the conclusion.

e. We now show $\chi_X \in \mathcal{F}$ if $|E| < \infty$, any set of finite measure. Write $E = H \setminus Z$. H is G_δ , $|Z| = 0$. Then $\chi_E = \chi_H \setminus \chi_Z$ and use linearity.

f. integrable simple functions $\in \mathcal{F}$. Then nonnegative integrable in \mathcal{F} . Then integrable in \mathcal{F} since $f = f^+ - f^-$. \square

To use Fubini, we need $f \in L^1$. A lot of times, this is what we are trying to show! Fubini cannot be used to prove $f \in L^1(I_1 \times I_2)$. Take $f(x, y) = \frac{1}{\sqrt{xy}}$ on $I := [0, 1] \times [0, 1]$. Cannot say $\int_I f = \int_0^1 (\int_0^1 \frac{1}{\sqrt{xy}} dx) dy$ but need f integrable, which is what you are trying to show. For these purposes, we use Tonelli's Theorem.

Theorem 4.2. Same as Fubini Theorem, $f \geq 0$, measurable. Suppose $f \geq 0$, measurable on $I_1 \times I_2$. Then

a): For a.e. $x \in I_1$, the slice function $y \mapsto f(x, y)$ is measurable on I_2 . b): the once-integrated function $x \mapsto \int_{I_2} f(x, y) dy$ is measurable on I_1 and its integral is $\int_I f$ (could be ∞).

In the example above, $\int_0^1 \frac{1}{\sqrt{xy}} dy = \frac{2}{\sqrt{x}}$ is measurable and $\int_0^1 \frac{2}{\sqrt{x}} dx = 4$ so $\int_I \frac{1}{\sqrt{xy}} = 4$.

Proof. Let $f_k = \min(f, k)\chi_{[-k, k]^n}$ integrable and $f_k \nearrow f$. Then Fubini applies. Spelled out. $\int_{I_1 \times I_2} f_k = \int_{I_1} \left(\int_{I_2} f_k(x, y) dy \right) dx$. The inner integral converges a.e. y to $\int_{I_2} f(x, y) dy$ by MCT. Then left MCT $\int_I f$. And then MCT for outer integral. done. \square

4.2 Applications of Fubini's Theorem

Let $f: E \rightarrow [0, \infty]$ be measurable. $\omega_f(y) = |\{x \in E: f(x) > y\}|$. Then $\int_E f = \int_0^\infty \omega_f(y) dy$. *Proof.* Recall $R(f, E) = \{(x, y) \in \mathbb{R}^{n+1}: x \in E, 0 \leq y \leq f(x)\}$. $\Gamma(f, E) = \{(x, y) \in \mathbb{R}^{n+1}: y = f(x)\}$. Now $|\Gamma| = 0$, $g = \chi_{R \setminus \Gamma}$ so $g(x, y) = 1$ if $0 \leq y < f(x)$, 0 otherwise.

$\int(\int g(x,y) dy) dx = \int(\int g(x,y) dx) dy$. and $\int f(x)\chi_E dx = \int \omega_f(y) dy$ so $\int_E f = \int_0^\infty \omega_f(y) dy$. This last integral is even Riemann integrable since it is monotone.

Generalizations: f measurable, $\int_E |f|^p = p \int_0^\infty y^{p-1} \omega_{|f|}(y) dy$. $p > 0$.

We know $\int_E |f|^p = \int_0^\infty \omega_{|f|^p}(y) dy$. Let $y = z^p$. Then this is $p \int_0^\infty z^{p-1} \omega_{|f|^p}(z^p) dz = p \int_0^\infty z^{p-1} \omega_{|f|}(z) dz$ and $\{|f|^p > z^p\} = \{|f| > z\}$. \square

Ex: $f(x,y) = \frac{\sin xy}{(x^2 + y^2)^{3/2}} \in L^1(E)$. $E = \{(x,y) : x,y > 0\}$.

Idea: estimate above by nonnegative function g and prove $g \in L^1(E)$ by Tonelli. i.e. $|f| \geq g$. $|f| \geq \frac{1}{(x^2 + y^2)^{3/2}}$ for sure but problems around origin but good at infinity. Have $|f| \leq \frac{xy}{(x^2 + y^2)^{3/2}}$ good near 0. So use min of the two functions. Better: Use first when $x^2 + y^2 \leq 1$: $g_1 = \frac{1}{(x^2 + y^2)^{3/2}} \chi_{\{x^2 + y^2 \geq 1\}}$ and $\omega_{g_1}(\alpha) = \pi/(4\alpha^{2/3}) - \pi/4$ if $\alpha < 1$ and 0 otherwise. Note that $1/r^3 \cdot \chi_{r \geq 1}$ and $\alpha < 1$ $1/r^3 \alpha$ and $r = 1/\alpha^{1/3}$ and $1/4$ if circle roots $1/\alpha^{1/3}$.

Slicing a subset of \mathbb{R}^{n+m} . Given a measurable set $E \subset \mathbb{R}^{n+m}$, consider $E_x = \{y \in \mathbb{R}^m : (x,y) \in E\}$.

- E_x is measurable for almost all x .
- If $|E| < \infty$, then $|E_x| < \infty$ for almost every x .
- If $|E| = 0$, then $|E_x| = 0$ for almost every x .

All these things follow easily from applying Tonelli to χ_E since almost every slice function is measurable. Slice fo χ_E is χ_{E_x} and $\int_{\mathbb{R}^{n+m}} \chi_E < \infty$ then $\int_{\mathbb{R}^n} |E_x| dx < \infty$.

Example 4.2. Assuming the Continuum Hypothesis, there exists $E \subset [0,1]^2$ such that every vertical slice E_x is countable and every horizontal slice E_y is co-countable, i.e. $[0,1] \setminus E_y$ countable set. Using Tonelli

$$\int \left(\int \chi_E dy \right) dx = 0$$

since every vertical slice is countable. The other way

$$\int \left(\int \chi_E dx \right) dy = 1$$

This means E is not measurable in \mathbb{R}^2 (otherwise these would be equal). Then let $E = \{(x,y) : y \prec x\}$.

Products: If $E_1 \subset \mathbb{R}^n, E_2 \subset \mathbb{R}^m$, are measurable, then $E_1 \times E_2$ is measurable in \mathbb{R}^{n+m} and $|E_1 \times E_2| = |E_1| |E_2|$, taking $0 \cdot \infty = 0$. Assume $E_1 \times E_2$ is measurable, then

$$|E_1 \times E_2| = \int \left(\int \chi_{E_1 \times E_2}(x, y) dy \right) dx = \int |E_2| \chi_{E_1}(x) dx.$$

It remains to prove that $E_1 \times E_2$ is measurable. It suffices to prove if when $|E_1| < \infty$ and $|E_2| < \infty$. In the general case, $E_1 = \bigcup_{k=1}^{\infty} E_{1,k}$ and $E_2 = \bigcup_{k=1}^{\infty} E_{2,k}$, then $E_1 \times E_2 = \bigcup_{j,k} E_{1,k} \times E_{2,j}$.

There exists $G_1 \supset E_1, G_2 \supset E_2$ such that $|G_1 \setminus E_1| < \epsilon, |G_2 \setminus E_2| < \epsilon$. Now $G_1 \times G_2$ is open and $E_1 \times E_2 \subset G_1 \times G_2$. Now $(G_1 \times G_2) \setminus (E_1 \times E_2) \subset [(G_1 \setminus E_1) \times G_2] \cup [G_1 \times (G_2 \setminus E_2)]$. Both $G_1 \setminus E_1$ and $G_2 \setminus E_2$ have measure at most ϵ . Now G_1, G_2 by cumes I_k, Q_j with summed measure at most ϵ . Then $|I_k \times Q_j| = |I_k| |Q_j|$ so $|(G_1 \setminus E_1) \times G_2| < \sum_{k,j} |I_k| |Q_j| < \epsilon |G_2|$. Then just choose ϵ sufficiently small.

Convolution: $(f * g)(x) = \int_{\mathbb{R}^n} f(x-t)g(t) dt$, where f, g are defined on \mathbb{R}^n are measurable. Now $f * g = g * f$ by change of variable $s = x - t$ so $\int f(x-t)g(t) dt = \int f(s)g(x-s) ds$, where $s, t \in \mathbb{R}^n$. In the general case, $h = h^+ - h^-$.

Claim $\int h(x-t) dt = \int h(t) dt$. **Proof:** if $h \geq 0$, then are measures of sets $\{(t, y) : 0 \leq y \leq h(x-t)\}$ and $\{(t, y) : 0 \leq y \leq h(t)\}$. The first is obtained from the other by reflection $t \mapsto -t$ and translation (add x). These do not change measure or measurability.

Example 4.3. Ex: Convolution: Argument of $\chi_{[a,b]} * \chi_{[c,d]}$ is only nonzero on $\{x : a \leq x-t \leq b, c \leq t \leq d\}$ so $\{x : x-b \leq t \leq x-a, c \leq t \leq d\}$. so on $\max(x-b, c) \leq t \leq \min(x-a, d)$.

$$\chi_{[a,b]} * \chi_{[c,d]} = \int_{\mathbb{R}} \chi_{[a,b]}(x-t) \chi_{[c,d]}(t) dt = (\min(x-a, d) - \max(x-b, c))^+$$

So $\chi_{[-1,1]} * \chi_{[-1,1]} = (\min(x+1, 1) - \max(x-1, -1))^+ = \dots$ grossly complicated (or just wrong). So better way $|([-b, -a] + x) \cap [c, d]|$ so $|([-1, 1] + x) \cap [-1, 1]|$. Then get

Theorem 4.3. If $f, g \in L^1(\mathbb{R}^n)$, then $f * g \in L^1(\mathbb{R}^n)$ and $\int |f * g| \leq \int |f| \int |g|$.

Proof. We know $|f * g| \leq |f| * |g|$ by the triangle inequality. We can work with $|f|$ and $|g|$ so assume $f, g \geq 0$. $\int (\int f(x-t)g(t) dt) dx$ by tonelli this is $\int (\int f(x-t)g(t) dx) dt = \int g(t) (\int f(x-t) dx) dt$ where the inner integral is just $\int f$ by translation. But then this final integral is $\int f \cdot \int g$.

But why $f(x-t)g(t)$ measurable on \mathbb{R}^{2n} ? (need this for tonelli). Observe $f(x)$ and $g(t)$ are measurable on \mathbb{R}^{2n} since the preimages of $> \text{something}$ is just $\{(x, t) : f > a\} = \{x : f(x) > a\} \times \mathbb{R}$ since it does not depend on t . Then this is measurable. Use the linear transformation $\phi : (x, t) \rightarrow (x-t, x+t)$ invertible linear, hence lipschitz. Use then HW 3.5: composition $f \circ \phi$ is measurable, ϕ^{-1} Lipschitz. Hence $f(x-t)$ is measurable on \mathbb{R}^{2n} .

If f, g are probability density functions of X and Y , then $f * g$ is the PDF of $X + Y$, assuming X, Y are independent.

4.3 Marcinkiewicz Integrals

In Analysis, one often has $f * g$, where one of the functions is not in L^1 . These are so called “singular integrals.” The simplest example is the Hilbert transform

$$\int_{\mathbb{R}} \frac{f(x-t)}{t} dt$$

Let K be a compact set in \mathbb{R}^n , and $\delta(y) := \text{dist}(y, K) = \min\{|x - y| : x \in K\}$.

Theorem 4.4. *If $B \supset K$ is a bounded open set, then for all $\lambda > 0$.*

$$\int \frac{\delta(y)^\lambda}{|x - y|^{n+\lambda}} dy < \infty$$

for a.e. $x \in K$.

Proof. It suffices to show $\int_K \mu_\lambda(x) dx < \infty$.

$$\int_K \mu_\lambda(x) dx = \int_K \int_{B \setminus K} \frac{\delta(y)^\lambda}{|x - y|^{n+\lambda}} dy dx = \int_{B \setminus K} \int_K \frac{\delta(y)^\lambda}{|x - y|^{n+\lambda}} dx dy.$$

But now observe $|x - y| \geq \delta(y)$. But then

$$\int_{B \setminus K} \int_K \frac{\delta(y)^\lambda}{|x - y|^{n+\lambda}} dx dy \leq \int_{B \setminus K} \left(\int_{|x-y| \geq \delta(y)} \frac{\delta(y)^\lambda}{|x - y|^{n+\lambda}} \right) dy \leq \int_{B \setminus K} C dx = C|B \setminus K| < \infty.$$

Using $\int_{|x-y| \geq \delta} \frac{1}{|x - y|^{n+\lambda}} dx \leq \frac{C}{\delta^\lambda}$. To prove this:

Recall $\int |f|^p = p \int_0^\infty \alpha^{p-1} |\{|f| > \alpha\}| d\alpha = *$. Use $P = n + \lambda$, $f = \frac{1}{|x - y|}$.

$$|\{|x - y| \geq \delta : \frac{1}{|x - y|} > \alpha\}| = |B(y, 1/\alpha) \setminus B(y, \delta)| = C((1/\alpha)^n - \delta^n)^+$$

which is 0 if $\alpha > 1/\delta$. For $*$, note

$$* = (n + \lambda)^{1/\delta} \alpha^{n+\lambda-1} C(1/\alpha^n - \delta^n) d\alpha \leq C' \int_0^{1/\delta} \alpha^{\lambda-1} d\alpha = C'' \frac{1}{\delta^\lambda}$$

B is needed as the integral is $\sim \frac{1}{|y|^n}$ as $y \rightarrow \infty$. We could have integrate over $B \setminus K$ since $\delta \equiv 0$ on K . Also, the δ is important for convergence. The almost always is also unavoidable: take $K = \{0\}$ in \mathbb{R} . Then $\delta(y) = |y|$. Then $\int_B \frac{|y|^\lambda}{|x - y|^{n+\lambda}} dy$ so at $x = 0$, we get $\int_B \frac{1}{|y|^n} = \infty$. Generally, $\mu_\lambda(x) = \infty$ if x is an isolated point of K . If x is an interior point fo K , $\mu_\lambda(x) < \infty$ as $\delta = 0$ in a neighborhood of x .

Question: in last part of proof, can we let $\lambda = 0$? Specifically, is it true that $\int_{B \setminus K} \frac{1}{|x - y|^n} dy < \infty$ for a.e. $x \in K$? Counterexample? K is fat Cantor set?

5 L^p Spaces

5.1 Introduction

Recall that a metric space is called complete if every Cauchy sequence converges. While most metric spaces are not complete, every metric space can be embedded into a complete space; that is, given a metric space X , there exists a space \tilde{X} such that there is an injection $\iota : X \hookrightarrow \tilde{X}$. The space \tilde{X} can be taken to be the set of equivalence classes of Cauchy sequences in X . Then X naturally embeds into \tilde{X} as the constant sequences. In fact using this construction, X is a dense subset of \tilde{X} .

Example 5.1.

- (i) The metric spaces \mathbb{R}^n and \mathbb{C}^n are complete metric spaces. This essentially follows from the Bolzano-Weierstrass Theorem.
- (ii) The space $C[a, b]$ of continuous real-valued functions on $[a, b]$ is a Banach space, and hence is complete.
- (iii) \mathbb{Q} is not a complete metric space. For example, it is routine to verify the sequence $\{x_n\}$, where $x_0 = 1$ and $x_{n+1} = \frac{x_n^2 + 2}{2x_n}$, is Cauchy and that (using the Monotone Convergence Theorem) $\lim_{n \rightarrow \infty} x_n = \sqrt{2} \notin \mathbb{Q}$.
- (iv) The space of continuous functions on $[0, 1]$ with metric $d(f, g) = \int_0^1 |f - g| dx$ is not complete. [Here the integral should be taken as the Riemann integral.] To see this, take any sequence of functions converging to the Dirichlet function. More concretely, it is routine to verify that the sequence of functions $\{f_n\}$, where

$$f_n(x) := \begin{cases} 1, & x = 0 \\ 1 - nx, & 0 < x < \frac{1}{n} \\ 0, & \frac{1}{n} \leq x \end{cases}$$

is Cauchy. However, $\lim_{n \rightarrow \infty} f_n(x)$ is the zero function $[0, 1]$ except at $x = 0$, where the function is 1, which is not continuous. One can extend this example by considering the space of Riemann integrable functions on $[0, 1]$ with the same metric d . Consider the sequence of compact sets $\{F_k\}$ used in constructing the Fat Cantor Set F , c.f. Lemma 1.3. We know that χ_{F_k} is Riemann integrable for all k and that $\int |\chi_{F_k} - \chi_{F_j}| < \epsilon$ for all $k, j \geq N$, where N is taken sufficiently large. However, the function χ_F is not Riemann integrable. Therefore, $\lim \chi_{F_k}$ does not exist with respect to the given metric d .

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In any introductory course in Functional Analysis, a great deal of time is spent analyzing Banach spaces, which are a type of topological vector space.² These play roles in Analysis and the rest of Pure Mathematics, Statistics, Probability, Physics, Finance, Engineering, etc.. We want to build a theory of complete spaces based on integration. The spaces we will build, the “Lebesgue spaces” L^p , in doing so will serve as important examples in Functional Analysis.

5.2 Definition of L^p and Examples

Suppose that $E \subseteq \mathbb{C}^n$ or $E \subseteq \mathbb{R}^n$ and let $f : E \rightarrow \mathbb{C}$ or $f : E \rightarrow \mathbb{R}$. Then $f = (f_1, \dots, f_n)$ is measurable if and only if each of the functions f_1, \dots, f_n are measurable. Equivalently, the sets $\{a_j < f_j < b_j\} = f^{-1}(\prod_{j=1}^m (a_j, b_j))$ are measurable for all j and $a_j, b_j \in \mathbb{R}$ if and only if for every countable union of open boxes U , $f^{-1}(U)$ is measurable.

Definition ($L^p, 0 < p < \infty$). Let $E \subseteq \mathbb{R}^n$ be measurable, and $0 < p < \infty$. Then we define

$$L^p(E) := \left\{ f : \int_E |f|^p < \infty \right\}$$

Again, we allow f to be complex-valued. In this case, if we write $f = f_1 + if_2$ for measurable real-valued functions f_1, f_2 , we have $|f|^2 = f_1^2 + f_2^2$. But then $|f_1|, |f_2| \leq |f| \leq |f_1| + |f_2|$. It follows then that $f \in L^p(E)$ if and only if $f_1, f_2 \in L^p(E)$. We write

$$|f|_{p,E} := \left(\int_E |f|^p \right)^{1/p}, \quad 0 < p < \infty.$$

Therefore, $L^p(E)$ is the set of equivalence classes of measurable functions f for which $|f|_{p,E} < \infty$. Finally, whenever the set E is clear from context, we will write L^p instead of $L^p(E)$ and $|f|_p$ instead of $|f|_{p,E}$. This this notation, we have $L^1 = L$.

Remark. Strictly speaking, L^p is a set of equivalence classes of functions f such that $\int_E |f|^p < \infty$; that is, we say two functions f, g define the same equivalence class in $L^p(E)$ if and only if $f = g$ a.e. on E .

For $c \in \mathbb{R}$ or $c \in \mathbb{C}$ and $f \in L^p(E)$, it is clear that $c \cdot f \in L^p(E)$; that is, $c[f] = [cf]$. Furthermore, if $f, g \in L^p(E)$, then

$$|f + g|^p \leq (|f| + |g|)^p \leq [2 \max(|f|, |g|)]^p = 2^p \max(|f|^p, |g|^p) \leq 2^p(|f|^p + |g|^p)$$

so that $f + g \in L^p(E)$. This shows that L^p is an \mathbb{R} , respectively \mathbb{C} , vector space. To define $L^p(E)$ in the case where $p = \infty$, we will first need a definition.

²To be precise, a Banach space is a complete normed vector space.

Definition (Essential Supremum). Let f be a real or complex valued function on a measurable set E of positive measure. Then we define $\sup_E \text{ess } f$ as follows: if $|\{x \in E: f(x) > \alpha\}| > 0$ for all real α , then we define $\sup_E \text{ess } f = \infty$ and otherwise define $\sup_E \text{ess } f := \inf\{\alpha: |\{x \in E: f(x) > \alpha\}| = 0\}$.

The essential supremum is the generalization of the maximums to measurable functions; the difference being the fact that values of a function on sets of measure zero do not affect the essential supremum. As such, we say that f is essentially bounded (or simply bounded) if $\sup_E \text{ess } |f| < \infty$.

Example 5.2.

- (i) Take $E = \mathbb{R}$ and consider $f = \chi_{\mathbb{Q}}$. Then $\sup_{\mathbb{R}} \text{ess } f = 0$ since $|\mathbb{Q}| = 0$ and $f \equiv 0$ on $\mathbb{R} \setminus \mathbb{Q}$.
- (ii) Take $E := [0, 1]$, where F is the fat Cantor set, c.f. Lemma 1.3, and $f = \chi_F$. Then $\sup_E \text{ess } \chi_F = 1$ by similar reasoning to (i).
- (iii) Let $E = \mathbb{R}$ and define

$$f(x) = \begin{cases} -10, & x = -1 \\ 10, & x = 1 \\ \pi, & \text{otherwise} \end{cases}$$

Then $\sup_E \text{ess } f = \pi$. Notice that the largest value that f takes is 10 but it takes this value on a set of measure zero.

- (iv) Let $E = \mathbb{R}$ and define

$$f(x) = \begin{cases} x^3 + x + 1, & x \in \mathbb{Q} \\ \arctan x, & x \in \mathbb{R} \setminus \mathbb{Q} \end{cases}$$

As in (i), since $|\mathbb{Q}| = 0$, we have $\sup_E \text{ess } f = \frac{\pi}{2}$.

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Definition ($L^\infty(E)$). Let $E \subseteq \mathbb{R}^n$ be a measurable set of positive measure. Then we define

$$L^\infty(E) = \{\text{meas } f : E \rightarrow \mathbb{C} : \sup_E \text{ess } |f| < \infty\}$$

First, note we write $|f|_\infty = |f|_{\infty, E} = \sup_E \text{ess } |f|$. The fact that $f \in L^\infty(E)$ is equivalent to the fact that there exists $M \in \mathbb{R}$ such that $|f| \leq M$ a.e.. Thus, $L^\infty(E)$ is the set of functions which are essentially bounded on E .

Example 5.3.

(i) Let $E = \mathbb{R}$ and consider the function $f = \frac{1}{|x|}$. The function f is a.e. finite on E but it is not bounded so that $f \notin L^\infty$.

(ii) Consider the function $f(x) = \frac{1}{\sqrt{x}}$ on the set $E = (0, 1)$. Then $f \in L^p$ if

$$\int_{(0,1)} \left(\frac{1}{\sqrt{x}} \right)^p < \infty.$$

But this occurs if and only if $p/2 < 1$. Therefore, $f \in L^p$ if and only if $p < 2$.

(iii) Let $E = [1, \infty)$ and define $f = \frac{1}{\sqrt{x}}$. Then $f \in L^p$ if and only if

$$\int_{[1,\infty)} \frac{1}{x^{p/2}} dx.$$

But this happens if and only if $p > 2$ or $p = \infty$. Therefore, $f \in L^p$ if and only if $p \in [2, \infty]$.

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5.3 Relations between L^p spaces

Not all the L^p spaces are ‘created equally’—there are ‘nicer’ and ‘uglier’ L^p spaces. Ranked, the L^p spaces are as follows:

1. (Nicest) L^2
2. $L^p, 1 < p < \infty, p \neq 2$
3. L^1
4. L^∞
5. (Worst) $L^p, 0 < p < 1$.

The reasoning for placing L^2 first is essentially because it is probably the most familiar to the reader. The reasoning for this ordering will make itself apparent in this section. The simplest relationship is between L^∞ and L^p for $p < \infty$.

Proposition 5.1. *If $|E| < \infty$, then $|f|_\infty = \lim_{p \rightarrow \infty} |f|_p$.*

Proof. Let $M = |f|_\infty$. If $M' < M$, then the set $A = \{x \in E: |f(x)| > M'\}$ must have positive measure. Furthermore,

$$|f|_p \geq \left(\int_A |f|^p \right)^{1/p} \geq M'|A|^{1/p}.$$

However, $|A|^{1/p} \rightarrow 0$ as $p \rightarrow \infty$. This proves that $\liminf_{p \rightarrow \infty} |f|_p \geq M'$. Therefore, $\liminf_{p \rightarrow \infty} |f|_p \geq M$. But we also have

$$|f|_p \leq \left(\int_E M^p \right)^{1/p} = M|E|^{1/p}.$$

Therefore, $\limsup_{p \rightarrow \infty} |f|_p \leq M$. □

Note that the requirement that $|E| < \infty$ is necessary: the constant function $f(x) = c$, $c \neq 0$, is in L^∞ but is not in L^p for any $0 < p < \infty$. Now we show that the L^p spaces are nested, and in a sense, form an interval.

Proposition 5.2. *If $|E| < \infty$, and $f \in L^{p_2}$ then $f \in L^{p_1}$ for all $0 < p_1 < p_2$. Furthermore, if $f \in L^{p_1}$ and $f \in L^{p_2}$, then $f \in L^p$ for all $p_1 < p < p_2$. That is, $\{p: f \in L^p\}$ is an interval starting from 0.*

Proof. Let $f \in L^{p_2}$. Write $E = E_1 \cup E_2$, where $E_1 := \{x: |f| \leq 1\}$ and $E_2 := \{x: |f| > 1\}$. Then we have

$$\int_E |f|^p = \int_{E_1} |f|^p + \int_{E_2} |f|^p, \quad 0 < p < \infty.$$

Now we have $\int_{E_1} |f|^p \leq |E_1|$. The integral $\int_{E_2} |f|^p$ is an increasing function in p . As $f \in L^{p_2}$ and $p_1 < \infty$, then $f \in L^{p_1}$. If $p_2 = \infty$, then f is a bounded function on a set of finite measure so that $f \in L^{p_1}$. For the second statement, suppose that $p_2 < \infty$. We have

$$\begin{aligned} \int_{E_1} |f|^p &\leq \int_{E_1} |f|^{p_1} < \infty \\ \int_{E_2} |f|^p &\leq \int_{E_2} |f|^{p_2} < \infty \end{aligned}$$

The result then follows. Now suppose that $p_2 = \infty$, and let $M := \sup_E \text{ess } |f|$. Proving $f \in L^p$ except $p_2 = \infty$ is a special case. let $M = \sup_E \text{ess } |f|$. The integral $\int_{E_1} |f|^p = 0$. It suffices to show the integral $\int_{E_2} |f|^p$ is finite. We have

$$\int_{E_2} |f|^p \leq \int_{E_2} M^p \leq M^p |E_2| \leq M^p |\{ |f|^{p_2} > 1 \}| \leq M^p \int_{|f|^{p_2} > 1} |f|^{p_1} < \infty$$

□

Again, the requirement that $|E| < \infty$ is necessary. For example, $f(x) = x^{-1/p_1} \in L^{p_2}([1, \infty))$ for $p_1 < p_2$ but $f(x) \notin L^{p_1}([1, \infty))$. Moreover, any nonzero constant is in L^∞ but not in $L^{p_1}(E)$ if $|E| = \infty$ and $p_1 < \infty$. A function may belong to L^{p_1} for $p_1 < p_2$ and not belong to L^{p_2} . For example, if $p_2 < \infty$ then $f(x) = x^{-1/p_2} \in L^{p_1}((0, 1))$, where $p_1 < p_2$, but $f(x) \notin L^{p_2}((0, 1))$. Furthermore, $f(x) = -\log x \in L^{p_1}((0, 1))$ for $p_1 < \infty$ but $f(x) \notin L^\infty((0, 1))$. For any set E (even with infinite measure), if $f \in L^{p_1}(E)$, then $f \in L^{p_2}(E)$, where $p_1 < p_2$. Finally, there are functions which are not in L^p for any p . For example, if $f \equiv 1$ on \mathbb{R} , then $f \notin L^p$ for all p except for $p = \infty$. If $f(x) = x$ on \mathbb{R} , then $f \notin L^p$ for all $p \in (0, \infty]$. One can find examples on finite intervals also. As a final remark for this section, if $|E| < \infty$, we have

$$\sup_E \text{ess } |f| = \lim_{p \rightarrow \infty} \left(\int_E |f|^p \right)^{1/p}$$

We are then able to write

$$\|f\|_{p,E} = \begin{cases} \left(\int_E |f|^p \right)^{1/p}, & 0 < p < \infty \\ \sup_E \text{ess } |f|, & 0 < p < 1 \end{cases}$$

5.4 Hölder and Minkowski Inequalities

In this section, we prove two important and powerful inequalities: Hölder's and Minkowski's Inequality. Hölder's inequality is about controlling products while Minkowski is about controlling sums. These are among the most useful tools in Analysis. We begin with a more basic inequality.

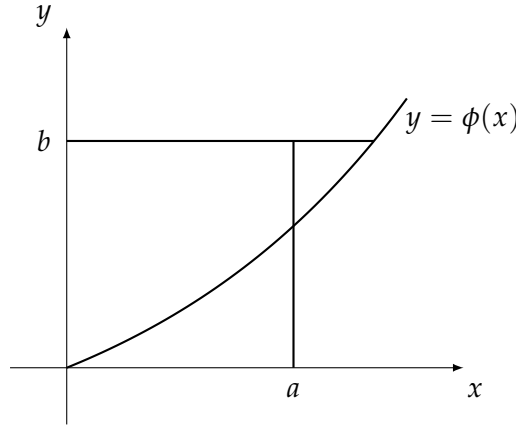
Theorem 5.1 (Young's Inequality). *Let $y = \phi(x)$ be continuous, real-valued, and strictly increasing for $x \geq 0$ with $\phi(0) = 0$. If $x = \psi(y)$ is the inverse of ϕ , then for $a, b > 0$,*

$$ab \leq \int_0^a \phi(x) dx + \int_0^b \psi(y) dy.$$

with equality holds if and only if $b = \phi(a)$.

Proof. The proof becomes immediate if we interpret each term as an area, and that the graph of ϕ may also serve as the graph of ψ by interchange of the x and y axes. Equality

holds if and only if (a, b) lies on the graph of ϕ .



□

Now taking $\phi(x) = x^\alpha$ for $\alpha > 0$, the $\psi(y) = y^{1/\alpha}$, and Young's Inequality is

$$ab \leq \frac{a^{1+\alpha}}{1+\alpha} + \frac{b^{1+1/\alpha}}{1+1/\alpha},$$

Setting $p = \alpha + 1$ and $p' = 1 + 1/\alpha$, we have

$$ab \leq \frac{a^p}{p} + \frac{b^{p'}}{p'},$$

where $a, b \geq 0$, $1 < p < \infty$, and $\frac{1}{p} + \frac{1}{p'} = 1$.

Definition (Conjugate Exponents). Numbers $p, p' > 0$ with $1/p + 1/p' = 1$ are called conjugate exponents. We make the convention that $1/\infty = 0$ here.

With this terminology, the numbers 1 and ∞ are conjugate. Other examples are the pairs $\{1.1, 11\}$, $\{1.5, 3\}$, and $\{2, 2\}$, which is the only number which is self-conjugate.

Axioms of a norm: $\|f\| \geq 0$, $\|f\| = 0$ if and only if $f = 0$: Positivity $\|\lambda f\| = |\lambda| \|f\|$ for $\lambda \in \mathbb{C}$ or \mathbb{R} homogeneity $\|f + g\| \leq \|f\| + \|g\|$ subadditivity (triangle inequality)

So $\|f\|_p$ is a norm when $1 \leq p \leq \infty$ (and not if $0 < p < 1$).

For $p = 1$ and $p = \infty$, this is easy: $p = 1$: $\int |f + g| \leq \int (|f| + |g|) = \int |f| + \int |g|$. For $p = \infty$, $|f| \leq M$ a.e. $|g| \leq N$ a.e., then $|f + g| \leq M + N$ a.e.. For $1 < p < \infty$, we need Hölder's inequality.

Theorem 5.2 (Hölder's Inequality). *If $1 \leq p \leq \infty$ and $1/p + 1/p' = 1$, then $\|fg\|_1 \leq \|f\|_p \|g\|_{p'}$, i.e.*

$$\begin{aligned} \int_E |fg| &\leq \left(\int_E |f|^p \right)^{1/p} \left(\int_E |g|^{p'} \right)^{1/p'}, \quad 1 < p < \infty \\ \int_E |fg| &\leq (\sup_E \text{ess } |f|) \int_E |g|. \end{aligned}$$

Proof. The last inequality is the case when $p = \infty$, and is routine. Suppose then that $1 < p < \infty$. If $\|f\|_p = \|g\|_{p'} = 1$, then

$$\int_E |fg| \leq \int_E \left(\frac{|f|^p}{p} + \frac{|g|^{p'}}{p'} \right) = \frac{\|f\|_p^p}{p} + \frac{\|g\|_{p'}^{p'}}{p'} = \frac{1}{p} + \frac{1}{p'} = 1 = \|f\|_p \|g\|_{p'}.$$

If $\|f\|_p$ or $\|g\|_{p'}$ is zero, then fg is zero a.e. in E and the result follows. We may then assume that neither $\|f\|_p$ nor $\|g\|_{p'}$ is zero. Setting

$$\tilde{f} = \frac{f}{\|f\|_p} \quad \text{and} \quad \tilde{g} = \frac{g}{\|g\|_{p'}}$$

we have $\|\tilde{f}\|_p = \|\tilde{g}\|_{p'} = 1$. But then in this case, the result follows by the work above. \square

Notice outer exponents must always add to 1 in Hölder's Inequality. The case where $p = 2$ is a classical result:

Corollary 5.1 (Cauchy-Schwarz Inequality).

$$\int_E |fg| \leq \left(\int_E |f|^2 \right)^{1/2} \left(\int_E |g|^2 \right)^{1/2}$$

Proof. This follows immediately from Hölder's Inequality using the self conjugate exponents $p = p' = 2$. \square

Example 5.4. Suppose $f \in L^6([1, \infty])$. Want to estimate the integral $\int_1^\infty \frac{f(x)^2}{x} dx$. Writing

$$\int_1^\infty \frac{f(x)^2}{x} dx = \int_1^\infty f(x)^2 \cdot \frac{1}{x} dx$$

We want to find a power p' such that the integral $1/x^{p'}$ converges. However, we still want to make use of the given information. We have f^2 and $f \in L^6$. Writing $f^6 = (f^2)^3$, we see

that $p' = 3/2$ will work. So we have

$$\begin{aligned} \int_1^\infty \frac{f(x)^2}{x} dx &= \int f(x)^2 \cdot \frac{1}{x} dx \leq \left(\int_1^\infty (f^2)^3 \right)^{1/3} \left(\int_1^\infty \frac{1}{x^{3/2}} \right)^{2/3} \\ &= \left(\left(\int_1^\infty f^6 \right)^{1/6} \right)^2 \cdot 2^{2/3} \\ &= 2^{2/3} \|f\|_6^2. \end{aligned}$$

◁

There is a sort of partial converse to Hölder's Inequality, which gives a dual form of the L^p norm.

Theorem 5.3. *Let f be real-valued and measurable on E , and $1 \leq p \leq \infty$ with conjugate exponent pair p' . Then*

$$\|f\|_p = \sup \int_E fg,$$

where the supremum is taken over all real-valued g with $\|g\|_{p'} \leq 1$ and $\int_E fg$ exists; that is,

$$\|f\|_p = \sup \left\{ \int_E fg : \|g\|_{p'} \leq 1, g \text{ real} \right\} = \sup \left\{ \frac{\int_E fg}{\|g\|_{p'}} : g \in L^{p'}, g \text{ real} \right\}.$$

Proof. The fact that the left side is larger than the right follows from Hölder's Inequality. It remains to prove the reverse inequality. If $\|f\|_p = 0$, then $f \equiv 0$ a.e. on E and the result follows. If $0 < \|f\|_p < \infty$, by dividing by $\|f\|_p$, we may assume that $\|f\|_p = 1$. One routinely verifies that $\|g\|_{p'} = 1$ and $\int_E fg = 1$, which completes this case.

Now suppose that $\|f\|_p = \infty$. Define f_k on E via

$$f_k(\mathbf{x}) = \begin{cases} \min(f(\mathbf{x}), k), & |\mathbf{x}| \leq k \\ 0, & |\mathbf{x}| > k \end{cases}$$

Now for all k , $f_k \in L^p$ and $\|f_k\|_p \rightarrow \|f\|_p = \infty$. By the work above, we have $\|f_k\|_p = \int_E f_k g_k$ for some $g_k \geq 0$ with $\|g_k\|_{p'} = 1$. But $f \geq f_k$ so that

$$\int_E fg_k \geq \int_E f_k g_k \rightarrow \infty.$$

But it then follows that

$$\sup_{\|g\|_{p'}=1} \int_E fg = \infty = \|f\|_{p'}.$$

Now if f is allowed to be nonnegative, apply the work above to $|f|$. Then there exists g_k with $\|g_k\|_{p'} = 1$ with

$$\|f\|_p = \lim \int_E |f| g_k = \lim \int_E f \tilde{g}_k,$$

where $\tilde{g} := (\text{sgn } f)g_k$. But then as $\|\tilde{g}_k\|_{p'} = 1$, the result follows. The cases $p = 1$ and $p = \infty$ are routine and left to the reader. \square

Furthermore, we have another sort of ‘reverse Hölder’ result. Suppose that for $f \in L^p$ and $g \in L^q$, we have $\int_{\mathbb{R}} |fg| \leq C\|f\|_p\|g\|_q$ for some C , then $p^{-1} + q^{-1} = 1$. To see this, rescaling to $f(\lambda x)$ and $g(\lambda x)$, we have

$$\int_{\mathbb{R}} f(\lambda x)g(\lambda x) dx = \frac{1}{\lambda} \int_{\mathbb{R}} fg.$$

Similarly,

$$\left(\int_{\mathbb{R}} |f(\lambda x)|^p \right)^{1/p} = \left(\frac{1}{\lambda} \right)^{1/p} \|f\|_p.$$

But then we have

$$\frac{1}{\lambda} \int |fg| \leq C \left(\frac{1}{\lambda} \right)^{1/p+1/q} \|f\|_p \|g\|_q.$$

However, this does not hold for all $\lambda > 0$ unless $p^{-1} + q^{-1} = 1$. Note also the special case in Hölder’s Inequality when $g \equiv 1$:

$$\int_E |f| = \int_E |f \cdot 1| \leq \left(\int_E \|f\|^p \right)^{1/p} \left(\int_E 1^{p'} \right)^{1/p'}.$$

But then

$$\int_E |f| \leq |E|^{1/p'} \|f\|_p.$$

This is vacuous if $|E| = \infty$.

Example 5.5. Let $1 < p \leq \infty$. Suppose $f : \mathbb{R} \rightarrow \mathbb{R}$ is C^1 and $f' \in L^p$. Using Hölder’s Inequality with $1/p'$, there exists C such that $|f(a) - f(b)| \leq C|a - b|^{1/p'}$. We then have

$$|f(a) - f(b)| \leq \int_a^b |f'(x)| dx \leq (b - a)^{1/p'} \|f'\|_p.$$

Notice that Lipschitz continuity arises in Hölder’s Inequality in the case where $p = \infty$, so that $p' = 1$. \triangleleft

Theorem 5.4 (Minkowski's Inequality). *If $1 \leq p \leq \infty$, then $\|f + g\|_p \leq \|f\|_p + \|g\|_p$, i.e.*

$$\left(\int_E |f + g|^p \right)^{1/p} \leq \left(\int_E |f|^p \right)^{1/p} + \left(\int_E |g|^p \right)^{1/p}$$

Proof. The proof in the case of $p = 1$ is immediate from the triangle inequality: $|f + g| \leq |f| + |g|$. Now if $p = \infty$, we know that $|f| \leq \|f\|_\infty$, $|g| \leq \|g\|_\infty$ a.e. in E . But then $|f + g| \leq \|f\|_\infty + \|g\|_\infty$ a.e. in E . It then follows that $\|f + g\|_\infty \leq \|f\|_\infty + \|g\|_\infty$. For $1 < p < \infty$, we have

$$\begin{aligned} \|f + g\|_p^p &= \int_E |f + g|^p \\ &= \int_E |f + g|^{p-1} |f + g| \\ &\leq \int_E |f + g|^{p-1} |f| + \int_E |f + g|^{p-1} |g|. \end{aligned}$$

Then using Hölder's Inequality, Theorem 5.2, to $|f + g|^{p-1}$, $|g|$, with exponents $p/(p-1)$, p , respectively, we have

$$\int_E |f + g|^{p-1} |g| \leq \|f + g\|_p^{p-1} \|g\|_p.$$

The same holds for the other integral mutatis mutandis. But then $\|f + g\|_p^p \leq \|f + g\|_p^{p-1} (\|f\|_p + \|g\|_p)$. Dividing by $\|f + g\|_p^{p-1}$, the result then follows. [Note that if $\|f + g\|_p = 0$, the result is immediate, and if $\|f + g\|_p = \infty$, either $\|f\|_p = \infty$ or $\|g\|_p = \infty$, the result is also immediate.] \square

Remark. Minkowski's Inequality fails for $0 < p < 1$. For example, take $E = (0, 1)$, and $f(x) = \chi_{(0, 1/2)}$, $g(x) = \chi_{(1/2, 1)}$. Then clearly $\|f + g\|_p = 1$ while $\|f\|_p + \|g\|_p = 2^{-1/p} + 2^{-1/p} = 2^{1-1/p} < 1$.

Minkowski's Inequality also works for infinite sums as well.

Theorem 5.5. *If $1 \leq p \leq \infty$, then $\|a_k + b_k\|_p \leq \|a_k\|_p + \|b_k\|_p$; that is,*

$$\left(\sum_k |a_k + b_k|^p \right)^{1/p} \leq \left(\sum_k |a_k|^p \right)^{1/p} + \left(\sum_k |b_k|^p \right)^{1/p}$$

For infinite series: if $\sum_k \|f_k\|_p < \infty$, then $\sum f_k \in L^p$ and $\|\sum f_k\|_p \leq \sum \|f_k\|_p$ for $1 \leq p \leq \infty$.

Proof. Minkowski then $\|\sum_{k=1}^N |f_k|\|_p \leq \sum_k \|f_k\|_p$ then MCT allows $N \rightarrow \infty$ here. So $\sum |f_k| \in L^p$. Hence, $\sum f_k \in L^p$.

5.5 Sequence Classes ℓ^p

Sequence spaces are closely related to Functional Analysis, among many other areas of mathematics. Perhaps the most important of these spaces are the ℓ^p spaces, which are special cases of L^p spaces for the counting measure on the natural numbers. These spaces are also important as they serve as classical examples of Banach spaces.

Definition (ℓ^p). If $\{a_k\}$ is a real (or complex sequence), for $0 < p < \infty$, define

$$\|a_k\|_p = \left(\sum_k |a_k|^p \right)^{1/p},$$

while for $p = \infty$, define $\|a\|_\infty = \sup_k |a_k|$. Then we define ℓ^p to be

$$\ell^p := \{ \{a_k\} : \|a_k\|_p < \infty \},$$

Proposition 5.3. If $\{a_k\}$ is a real (or complex sequence), then $\|a_p\|_p \leq \|a\|_q$ if $p \geq q$.

Proof. We can rescale so that $\|a\|_1 = 1$. Then $|a_k| \leq 1$ for all k . Hence, $\sum |a_k|^p \leq \sum |a_k|^q \leq 1$. So $\|a\|_q \leq 1$. If $p < \infty$. The case $p = \infty$ is easy. \square

A sequence $\{a_k\}$ can be represented by $f = \sum_{k=1}^{\infty} a_k \chi_{[k, k+1)}$. Then $\|f\|_p = \|a\|_p$ as $\int |f|^p = \sum_{k=1}^{\infty} \int_k^{k+1} |f|^p = \sum |a_k|^p$. Given sequence a, b , can apply Holder/mink to corresponding functions and get $\sum |a_k b_k| \leq \|a\|_p \|b\|_{p'}$ Holder $\|a + b\|_p \leq \|a\|_p + \|b\|_p$ Mink

Ex: $a_k = 1/k$ is in $\ell^p \iff p > 1$. More generally, $a_k = 1/k^r \in \ell^p \iff p > 1/r$, where $r > 0$.

For geometric sequences, $a_k = 1/2^k \in \ell^p \iff$, well for all $p \in (0, \infty]$.

$a_k = 1/(k^r \log^{2r}(k+1)) \iff p \geq 1/r$.

Norm: $\|x\| \geq 0$, $\|x\| = 0$ iff $x = 0$ $\|\lambda x\| = |\lambda| \|x\|$ $\|x + y\| \leq \|x\| + \|y\|$ Normed vector space: $(V, \|\cdot\|)$.

$1 \leq p \leq \infty$ ℓ^2 norm L^p L^p has norm $\|f\|_p = (\int |f|^p)^{1/p}$ ℓ^p has norm $(\sum |a_k|^p)^{1/p}$
 $p = \infty$, special.

Norms give metric which then have convergence, sequences, open sets, etc etc (all regular topology plus more, compact sets, close open convergence continuous functions, etc etc. So once have norm, do not need to define these terms as they come immediately from the metric immediately from the norm.

Now the main thing for today is the following: are these metric spaces complete?

Definition (Banach Space). A Banach Space is a complete normed linear space.

Complete means every Cauchy sequence converges.

Convergence in L^p : $f_k \rightarrow f$ means $\int |f_k - f|^p \rightarrow 0$ if $p < \infty$ or essential sup $|f_k - f| \rightarrow 0$ if $p = \infty$.

Example 5.6. Conversely, $k^{-1}\chi_{[0,k]} \rightarrow 0$ pointwise but not in L^1 . $k\chi_{[0,1/k]} \rightarrow 0$ a.e. but not in L^1 . The first converges in L^p for $1 < p \leq \infty$ while the second does not converge in any L^p for $1 \leq p \leq \infty$.

So convergence pointwise and convergence in L^p are not related for all p .

Lemma 5.1. If $f_k \rightarrow f$ in L^p , then $f_k \xrightarrow{m} f$ (convergence in measure).

Proof. $1 \leq p < \infty$: $|\{ |f_k - f| > \epsilon \}| = |\{ |f_k - f|^p > \epsilon^p \}| \leq 1/\epsilon^p \int |f_k - f|^p$ by Chebyshev and right integral $\rightarrow 0$ as $k \rightarrow \infty$. $p = \infty$: $f_k \rightarrow f$ in measure means $f_k \rightarrow f$ uniformly on $E \setminus Z$, $|Z| = 0$ hence in measure. \square

The converse false as previous example they converged in measure.

Theorem 5.6. L^p and ℓ^p , $1 \leq p \leq \infty$ are Banach spaces.

$1 \leq p < \infty$: Suppose $\{f_k\}$ is Cauchy in L^p . Then $\{f_k\}$ is Cauchy in measure. $|\{ |f_k - f_j| > \epsilon \}| \rightarrow 0$ as $k, j \rightarrow \infty$. By Lemma, From 4.4, we have there exists f such that $f_k \xrightarrow{m} f$. there exists subsequence $f_{k_i} \rightarrow f$ a.e. for a fixed i .

$$f_{k_i} - f_{k_j} \xrightarrow{a.e.} f_{k_i} - f \quad j \rightarrow \infty$$

By Fatou's Lemma:

$$\int |f_{k_i} - f|^p \leq \liminf_{j \rightarrow \infty} \int |f_{k_i} - f_{k_j}|^p$$

By cauchy. Conclude $\int |f_{k_i} - f|^p \leq \epsilon$ for sufficiently large i . So $f_{k_i} \rightarrow f$ in L^p . Hence $f_k \rightarrow f$. General fact: Cauchy sequence with convergent subsequence converges. \square

Prove separability: separable if has countable dense subset.

Theorem 5.7. L^p , ℓ^p are separable if $p < \infty$ not if $p = \infty$.

In $L^p(\mathbb{R}^n)$: $\{\sum_{fin} (a_k + ib_k)\chi_{Q_k} : a_k, b_k \in \mathbb{Q}, Q_k \text{ dyadic cube}\}$. is dense in L^p .

In $L^p(E)$, just restrict to E (for $p < \infty$).

In ℓ^p $\{(a_1, a_2, \dots, a_m, 0, 0, \dots)\}$ then real an imag. parts rational, $m \in \mathbb{N}$ is dense for $p < \infty$.

$p = \infty$ Find countably many elements f_α such that $\|f_\alpha - f_\beta\|_\infty = 1$ for all $\alpha \neq \beta$ in $L^\infty(\mathbb{R}^n)$: $\chi_{[0,a]^n}$ $a > 0$ in ℓ^∞ all $\{0, 1\}$ sequences. $L^\infty(E)$ in $L^\infty(E)$: $|E| > 0$ consider $\phi(a) = |E \cap [-a, a]^n|$, $a > 0$ is continuous.

$$\phi(b) - \phi(a) \leq |[-b, b]^n \setminus [-a, a]^n| = (lb)^n - (la)^n \rightarrow 0$$

An inner product on a vector space V : a function $\langle f, g \rangle \rightarrow \mathbb{C}$, $f, g \in V$. With $\langle g, f \rangle = \overline{\langle f, g \rangle}$ (conjugate symmetry). and $\langle \alpha f_1 + \beta f_2, g \rangle = \alpha \langle f_1, g \rangle + \beta \langle f_2, g \rangle$ (linear in first, hence second, argument). $\langle f, f \rangle = 0$ and 0 if and only if $f = 0$.

Example 5.7. $L^2(E)$ has $\langle f, g \rangle = \int_E f \bar{g}$. ℓ^2 has $\langle a, b \rangle = \sum_k a_k \bar{b}_k$. These are finite by Hölder, $p = p' = 2$.

General facts: $\|f\| = \sqrt{\langle f, f \rangle}$ is a norm $|\langle f, g \rangle| \leq \|f\| \|g\|$. (abstract Cauchy-Schwarz)
Hilbert space = complete inner product space.

L^2, ℓ^2 are Hilbert spaces.

Orthogonality: $f \perp g$ if $\langle f, g \rangle = 0$.

Pythagorean Theorem: if $f \perp g$, then $\|f \pm g\|^2 = \|f\|^2 + \|g\|^2$. Why? $\langle f + g, f + g \rangle = \langle f, f \rangle + \langle f, g \rangle + \langle g, f \rangle + \langle g, g \rangle = \|f\|^2 + \|g\|^2$.

More generally, if f_1, \dots, f_n are mutually orthogonal, then $\|\sum f_k\|^2 = \sum \|f_k\|^2$.

Orthonormal system: $\{\phi_k\}$ in Hilbert space such that

$$\langle \phi_k, \phi_j \rangle = \begin{cases} 0, & k \neq j \\ 1, & k = j \end{cases}$$

Example 5.8. In ℓ^2 , $\phi_1 = (1, 0, 0, \dots)$, $\phi_2 = (0, 1, 0, \dots)$, ... in L^2

Example 5.9. $\phi_k = e^{2\pi i k x}$ on $[0, 1]$ ($k \in \mathbb{N}$ or $k \in \mathbb{Z}$).

Why is it nice to have orthogonal systems? If $\{\phi_k\}$ is an ONS, then

$$\left\| \sum_{k=1}^n c_k \phi_k \right\|^2 = \sum_{k=1}^n |c_k|^2.$$

How do we let $n \rightarrow \infty$ here? If $c \in \ell^2$, i.e. $\sum_{k=1}^{\infty} |c_k|^2 < \infty$. Then $\sum_1^{\infty} c_k \phi_k$ converges in L^2 and $\|\sum c_k \phi_k\| = \sum_{k=1}^{\infty} (c_k)^2$.

Proof. We need for all ϵ such that there exists

$$\left\| \sum_{k=m}^n c_k \phi_k \right\| < \epsilon$$

whenever $m, n, \geq N$. But

$$\left\| \sum_{k=m}^n c_k \phi_k \right\| = \sqrt{\sum_{k=m}^n |c_k|^2}$$

which is small as $\sum |c_k|^2 < \infty$.

Also, let $s_n = \sum_1^n c_k \phi_k$, $s_n \rightarrow s = \sum_1^{\infty} c_k \phi_k$. $\|s_n\| - \|s_n - s\| \leq \|s\| \leq \|s_n\| + \|s_n - s\|$. On the right we have $\sqrt{\sum_1^n |c_k|^2}$ and a term tending to 0, same on the left. Then by Squeeze Theorem we have $\|s\| = \sqrt{\sum_1^{\infty} |c_k|^2}$.

Reverse engineering process from $f \in L^2$ to coefficients c_k . Fourier expansion. Note if $f = \sum c_k \phi_k$ then $c_k = \langle f, \phi_k \rangle$.

Why? $\langle f, \phi_j \rangle = \langle \sum c_k \phi_k, \phi_j \rangle = \sum c_k \langle \phi_k, \phi_j \rangle = c_j$. Why equality? For finite ONS, use linearity. For infinite, use continuity of $f \mapsto \langle f, g \rangle$, c.f. hw 8.4. or by previous result $\sum |c_k|^2 = \|f\|^2 < \infty$.

Linear span of a set E = set of all finite linear combinations of elements of E .

Theorem 5.8. Given $f \in L^2$ and $\{\phi_k\}$ a ONS in L^2 . TFAE

1) $f = \sum c_k \phi_k$ (series converges in L^2) 2) $f \in \overline{\text{span } \{\phi_k\}}$ (closure in L^2)

1 \rightarrow 2: if $f = \lim_{n \rightarrow \infty} \sum_1^n c_k \phi_k$, then $v_n = \sum^n c_k \phi_k$ is in the span since $v_n \rightarrow f$ in L^2 , we have $f \in \text{closure of the span}$.

2 \rightarrow 1: Let $c_k = \langle f, \phi_k \rangle$ and $s_n = \sum^n c_k \phi_k$. Want to show that $s_n \rightarrow f$. Let M be the span of $\{\phi_1, \dots, \phi_n\}$. $(s_n - f) \perp \text{span } (\phi_1, \dots, \phi_n)$. because $\langle s_n - f, \phi_k \rangle = c_k - c_k = 0$. Then use Pythagorean Theorem: for any g in the span of ϕ_1, \dots, ϕ_n , $\|f - g\| \geq \|f - s_n\|$. Now $\|f - g\|^2 = \|f - s_n\|^2 + \|s_n - g\|^2$. Since f is in the closure of the span up to ϕ_n the distance between f and the span tends to 0. So once n sufficiently large, close to 0. \square

Corollary 5.2. Since $f - s_n \perp \phi_1, \dots, \phi_n$, $\|f\|^2 = \sum^n |c_k|^2 + \|f - s_n\|^2$.

This has two consequences: Bessel Inequality: $\sum^\infty |c_k|^2 \leq \|f\|^2$. Parseval's Identity: $\sum |c_k|^2 = \|f\|^2$ if and only if $s_n \rightarrow f$.

6 Abstract Integration

6.1 Measure Spaces

So far, we only have had Lebesgue measure on \mathbb{R}^n . But there are other measures: length of curves in \mathbb{R}^2 , surface area on surfaces, probability on probability space Ω $P(\text{event}) = \text{measure of outcomes}$.

Definition (Measure Space). A measure space is a triple, (X, Σ, μ) , where X is a set, Σ is a σ -algebra on subsets of X , and $\mu : \Sigma \rightarrow [0, \infty]$ is a countably additive function on disjoint sets.

Remark. One often also requires $\mu(\emptyset) = 0$. However, this does almost follow from our definition since $\mu \geq 0$ and if $\emptyset \neq A \subseteq X$, then

$$\mu(A) = \mu(A \cup \emptyset) = \mu(A) + \mu(\emptyset)$$

so that $\mu(\emptyset) = 0$ unless $\mu(\emptyset) = \infty$.

Example 6.1.

- (i) $(\mathbb{R}^n, \mathcal{L}, \mu)$, \mathcal{L} is the set of Lebesgue measurable set. $\mu(E) = |E|$.
- (ii) $(X, 2^X, \mu)$, X is any set, $\mu(E)$ is the cardinality of the set.
- (iii) Like (ii) but

$$\mu(E) = \begin{cases} 0, & E \text{ countable} \\ \infty, & E \text{ uncountable} \end{cases}$$

- (iv) $(\mathbb{R}^n, \mathcal{L}, \mu_w)$, where

$$\mu_w(E) = \int_E w$$

where $w \geq 0$ is measurable. One may say that w is the ‘weight’ or ‘density.’

Non-Example 6.1. like (ii) above but

$$\mu(E) = \begin{cases} 0, & |E| < \infty \\ \infty, & |E| \text{ infinite} \end{cases}$$

this is *not* a measure space since it is not countably additive.

Definition (Additive Set Function (ASF)). We say a function $\phi : \Sigma \rightarrow \mathbb{R}$, where Σ is a σ -algebra, is countably additive.

The difference between ϕ and a measure is that it can be negative and it cannot be infinite. This is also called a *finite signed measure*.

Example 6.2. $\phi(E) = \int_E f$, where $f \in L^1$.

An advantage of ASF can pass to complement: $\phi(E^C) = \phi(X) - \phi(E)$ and get results for \cap from \cup .

Theorem 6.1. *ASF is continuous on monotone sequences of sets, i.e. $E_1 \subset E_2 \subset \dots$ then $\phi(\cup E_k) = \lim \phi(E_k)$ or $E_1 \supset E_2 \supset \dots$ then $\phi(\cap E_k) = \lim \phi(E_k)$*

Proof. The second follows from the first by taking complements. Now

$$\bigcup_{k=1}^{\infty} E_k = E_1 \cup (E_2 \setminus E_1) \cup (E_3 \setminus E_2 \setminus E_1) \cup \dots$$

disjoint union so

$$\phi(\cup E_k) = \phi(E_1) + (\phi(E_2) - \phi(E_1)) + \dots$$

which telescopes so we have $\lim_{k \rightarrow \infty} \phi(E_k)$. \square

For measures μ , the first works but in the second statement above, we require $\mu(E_1) < \infty$.

Theorem 6.2 ('Fatou's Lemma'). *if $\phi \geq 0$ is an ASF (i.e. finite measure), then $\phi(\liminf E_k) \leq \liminf \phi(E_k)$. Hence by considering E_k^C , $\phi(\limsup E_k) \geq \limsup \phi(E_k)$.*

Proof. Let $M_k = \bigcap_{j \geq k} E_j$, then $\liminf E_k = \bigcup_{k=1}^{\infty} M_k$.

$$\phi\left(\bigcup_{k=1}^{\infty} M_k\right) = \lim_{k \rightarrow \infty} \phi(M_k) = \liminf_{k \rightarrow \infty} \phi(E_k),$$

using $M_k \subset E_k$, so $\phi(E_k) - \phi(M_k) = \phi(E_k \setminus M_k) \geq 0$.

Jordan Decomposition for ASF

Every ASF is the difference of two nonnegative ASFs.

Application: Write ASF $\phi = \phi_1 - \phi_2$, both nonnegative. Conclude if $E = \limsup E_k = \liminf E_k$, then $\phi(E) = \lim \phi(E_k)$.

Definition (Upper Variation). Given an ASF ϕ , upper variation $\bar{V}(E) = \sup_{A \subset E} \phi(A)$.

Definition (Lower Variation). Given an ASF ϕ , lower variation $\underline{V}(E) = -\inf_{A \subset E} \phi(A)$.

Definition (Total Variation). Given an ASF ϕ , total variation: $V(E) = \bar{V}(E) + \underline{V}(E)$.

How does this relate to Jordan decomposition? Idea of proof is to show $\phi(E) = \bar{V}(E) - \underline{V}(E)$.

where (needs to be proved) \bar{V}, \underline{V} are ASFs. Why this was

$$\int_{A \subset E, A \in \Sigma} \phi(A) = \inf_{B \subset E, B \in \Sigma} (\phi(E) - \phi(B)) = \phi(E) - \sup_{B \subset E, B \in \Sigma} \phi(B) = \phi(E) - \bar{V}(E)$$

Take $A = E \setminus B$. Hence, $\phi(E) = \bar{V}(E) - \underline{V}(E)$.

Clear total variation is ASF if the upper/lower are. So sketch proof that they are:

1) Countable subadditivity. $A \subset \cup E_k$, write as $A = (A \cap E_1) \cup (A \cap E_2 \setminus E_1) \cup \dots$. Get $A = \cup A_k$, $A_k \subset E_k$. $\phi(A_k) \leq \bar{V}(E_k)$ hence $\phi(A) \leq \sum \bar{V}(E_k)$. The upper follows the same.

2) $V < \infty$, which is done by contradiction. $\bar{V}(E) \geq \sum \bar{V}(E_k)$ just pick $A_k \subset E_k$, $\phi(A_k) \geq \bar{V}(E_k) - \epsilon/2^k$, use countable subadditivity of ϕ .

6.2 Measurable Functions and Integrals

Repeat all previous work for abstract measure spaces. Recall (X, Σ, μ) is a measure space.

Definition. A function $f : X \rightarrow \bar{\mathbb{R}}$ is measurable if $\{f > a\} \in \Sigma$ for all $a \in \mathbb{R}$.

Fur functions $f : X \rightarrow \mathbb{R}$, equivalent to $f^{-1}(\text{Borel}) \in \Sigma$.

Example 6.3. Let $X = \mathbb{N}$, $\Sigma = 2^X$, then all functors are measurable.

Basic properties: cf , $f + g$, $\sup_k f_k$, $\inf_k f_k$, $\lim_k f_k$ are measurable.

$\sum_{\text{finite}} a_k \chi_{E_k}$, disjoint, a_k unique is measurable if and only if $E_k \in \Sigma$.

Every nonnegative measurable function f is the limit of f_k (simple), $f_k \nearrow f$, $f_k = \min(k, 2^{-k} \text{floor}(2^k f))$, second piece round down to $j/2^k$: $\{f_k = j/2^k\} = \{j/2^k \leq f < (j+1)/2^k\}$ or \emptyset .

What does not *not* generalize.

"continuous" then measurable. Continuity is not even defined on X .

Lusin's Theorem does not apply.

If f is measurable and $f = g$ a.e. wrt μ then g not necessarily measurable. (because subsets of measure zero no longer necessarily measurable). example: $(\mathbb{R}, \mathcal{B}, |\cdot|)$, Borel with Lebesgue measure. in Cantor set C , there is a non-Borel subset A . $\chi_A \equiv 0$ a.e. but χ_A is not measurable.

$X = \cup_{k=1}^{\infty} E_k$, E_k has finite measure. Example: \mathbb{R} with counting measure. ie ∞ on infinite sets and finite otherwise.

Definition. If $f \geq 0$ is measurable. $\int_E f d\mu = \sup\{\sum_j a_j \mu(E_j)\}$ over simple functions $\sum a_j \chi_{E_j} \leq f$.

Can choose $a_j = \inf_{E_j} f$, so $\int f d\mu = \sup_p L(f, P)$, where P is partition of E into E_1, \dots, E_m , $L(f, P) = \sum_j \int_{E_j} f d\mu$. Choosing finer partition P , get larger $L(f, P)$.

Definition. If f measurable and one of $\int_E f^+ d\mu$, $\int_E f^- d\mu$ is finite, define $\int_E f d\mu = \int_E f^+ d\mu - \int_E f^- d\mu$. if both finite, $f \in L^1(E, \mu)$.

Example 6.4. $(\mathbb{N}, 2^{\mathbb{N}}, \text{counting})$, $\int f d\mu = \sum_{k=1}^{\infty} f(k)$, if abs convergent. $L(f, P) = \sum_{k=1}^N f(k)$
So integrals are series! Notice integral finite iff abs convergent.

Example 6.5. $(\mathbb{R}, 2^{\mathbb{R}}, \delta_{x_0})$. δ_{x_0} the point mass at x_0 , i.e. dirac delta

$$\delta_{x_0} = \begin{cases} 1, & x_0 \in A \\ 0, & x_0 \notin A \end{cases}$$

$\int_{\mathbb{R}} f d\delta_{x_0} = f(x_0)$ *Proof.* Suppose f nonnegative (otherwise, use f^+ and $-f^-$ and then combine later). Let $P = \{E_1, \dots, E_m\}$ be a partition. One E_j contains x_0 , then $L(f, P) = \inf_{E_j} f \leq f(x_0)$. Hence, $\int f \leq f(x_0)$. Also, $P = \{\{x_0\}, \mathbb{R} \setminus \{x_0\}\}$ gives $L(f, P) = f(x_0)$.

We need another proof of MCT, since before used “volume definition” of $\int f$ and this does not generalize.

Lemma 6.1. *if $E = A \cup B$, disjoint measurable, then $\int_E f = \int_A f + \int_B f$ for nonnegative measurable f .*

Proof. Any partition P of E can be refined, (E_j replaced by $E_j \cap A, E_j \cap B$) so we get $P = P_A \cup P_B$. Then $L(f, P) = L(f_1, P_A) + L(f, P_B)$. Hence, $\int f \leq \int_A f + \int_B f$. But P_A, P_B approx $\int_A f, \int_B f$ part them into P , conclude $\int_E f \geq \int_A f + \int_B f$. \square

Example 6.6. $X = \mathbb{R}^2, \Sigma = \text{Borel sets}, \mu(E) = |\{x \in \mathbb{R}: (x, 0) \in E\}|$. $\int f d\mu = \int_{\mathbb{R}} f(x, 0) dx$. Important example of integrating along a curve.

Our goal is now to prove the MCT (if $f_k \nearrow f, f_k \geq 0$ measurable, then $\int f_k \rightarrow \int f$). We will need some preparation.

Subadditivity over f : $\int(f + g) = \int f + \int g$, where $f, g \geq 0$ measurable. How do we prove this? For simple, easy: $f = \sum a_k \chi_{E_k}, g = \sum b_l \chi_{F_l}$, rewrite as $\sum_{k,l} a_k \chi_{E_k \cap F_l}$ and $\sum_{k,l} b_l \chi_{E_k \cap F_l}$. The first is integral $\sum a_k \mu(E_k \cap F_l)$ and second is integral $\sum b_l \mu(E_k \cap F_l)$. But then for the sum: $\sum(a_k + b_l) \chi_{E_k \cap F_l}$ converges to integral $\sum(a_k + b_l) \mu(E_k \cap F_l)$.

Lemma 6.2. *Suppose $f_k, g \geq 0$ are simple, $f_1 \leq f_2 \leq \dots \lim f_k \geq g$. Then $\lim \int f_k \geq \int g$.*

Proof. Partition set E to make g constant (on each piece) $g \equiv c$. Given $c' < c$. Let $A_k = \{f_k > c'\}$. Then $A_1 \subset A_2 \subset \dots$ and $\bigcup A_k = E$, so $\mu(A_k) \rightarrow \mu(E)$. But $\int_E f_k \geq \int_{A_k} f_k \geq c' \mu(A_k)$. So $\lim \int f_k \geq c' \mu(E)$. So $\lim \int f_k \geq c \mu(E) = \int g$. \square

Lemma 6.3. *If $f_k \nearrow f, f_k$ simple nonnegative, then $\int f_k \rightarrow \int f$.*

Proof. For all simple $g \leq f$, we have $\lim \int f_k \geq \int g$ by the previous lemma. Take sup over g , get $\lim \int f_k \geq \int f$. But reverse is immediate since $f_k \nearrow f$ so $f_k \leq f$. \square

Proof of $\int(f + g) = \int f + \int g$.

Take simple $f_k \nearrow f, g_k \nearrow g$. Pass to limit in $\int(f_k + g_k) = \int f_k + \int g_k$. \square

Then follows for general function by writing f^+ and f^- .

Monotone Convergence Theorem (Two forms):

Sequence Form: if f_k are nonnegative, measurable, and $f_k \nearrow f$, then $\int f_k \rightarrow \int f$.

Series Form: If $g_k \geq 0$ measurable, then $\int \sum g_k = \sum \int g_k$.

These are equivalent. [The duality of sequences and series in a sense.]

Proof. (Series Form since easier to prove) $\int g \geq \sum_{k=1}^{\infty} \int g_k$ by considering $g \geq \sum_{k=1}^N g_k$. $\int g \geq \sum_{k=1}^N \int g_k$. Need \leq . Introduce simple $g_{k,j} \nearrow g_k$ as $j \rightarrow \infty$.

$$\begin{array}{ccccccc} g_{11} & g_{12} & g_{13} & \cdots & \rightarrow & g_1 \\ g_{21} & g_{22} & g_{23} & \cdots & \rightarrow & g_2 \\ g_{31} & g_{32} & g_{33} & \cdots & \rightarrow & g_3 \end{array}$$

We use an “upper triangular sum”. $s_j = \sum_{k=1}^j g_{k,j}$ which converges monotonically to g , i.e. $s_j \nearrow g$. By second lemma above, $\int s_j \rightarrow \int g$. Also, $\sum_{k=1}^j g_k \geq s_j$, hence $\sum_{k=1}^{\infty} \int g_k \geq \int g$. \square

MCT implies Fatou which implies DCT.

Hölder, Minkowski: $L^p(E, \mu)$ is a Banach space.

6.3 Absolute Continuity & Singular Sets

Now we move on to 10.3

Given a measure space (X, Σ, μ) and an ASF ϕ , we say

Definition (Absolutely Continuous). ϕ is absolutely continuous with respect to μ if $\mu(A) = 0$, then $\phi(A) = 0$.

Definition (Singular). ϕ is singular with respect to μ if there exists Z such that $\mu(Z) = 0$ and $\phi(A) = 0$ for all $A \subset Z^C$.

Equivalently, for total variation V

AC: $\mu(A) = 0$ then $V(A) = 0$

S: $\mu(Z) = 0$ and $V(Z^C) = 0$.

Main results: every ASF is the sum of AC and S. AC ASF has density $\phi(A) = \int_A \omega$.

Examples:

a) μ = Lebesgue measure on \mathbb{R} and $\phi(A) = \int_A f$, $f \in L^1(\mathbb{R})$. This is AC

b) same μ , $\phi(A)$ is 1 if $x_0 \in A$ and 0 if $x_0 \notin A = \delta_{x_0}$.

Singular: let $Z = \{x_0\}$ then $\mu(Z) = 0$, $V(Z^C) = 0$.

On $[0, 1]$, fix $w \geq 0$ integrable. fix $w \geq 0$, $w \in L$, $\mu_w(A) = \int_A w$.

Is Lebesgue measure on $[0, 1]$ AC wrt μ_w ? No if $|\{w = 0\}| > 0$

Yes if $w > 0$ a.e.

Goal: Every ASF is AC + S: AC has ‘density’ f , $\phi(A) = \int_A f$.

Remark: for any $E \subset \Sigma$, $(E, \Sigma' = \{F \in \Sigma : F \subset E\}, \mu|_{\Sigma'})$ is a measure space (subspace of X).

Theorem 6.3. (From $0-0$ to $\epsilon-\delta$ in AC) ϕ is AC if and only if for all $\epsilon > 0$, there exists $\delta > 0$ such that for all A $\mu(A) < \delta$ then $|\phi(A)| < \epsilon$.

Proof. Recall $\phi = \bar{V} - \underline{V}$, so $|\phi(A)| \leq V(A)$. Suppose there exists $\epsilon > 0$ for all k there exists A_k with $\mu(A_k) < 2^{-k}$ but $V(A_k) \geq \epsilon$. Let $A = \limsup A_k$. Then $\mu(A) = 0$ (because $\sum \mu(A_k) < \infty$) but $V(A) \geq \limsup V(A_k) \geq \epsilon$. (flipped Fatou). This contradicts AC. \square

Ex: For infinite measures μ, ν , this is not true. For example, μ is the Lebesgue measure on \mathbb{R} and $\nu(A) = \int_A x^2 dx$. ν is AC wrt μ . If $\mu(A) = 0$, then $\nu(A) = 0$ but for all $\delta > 0$, $\int_{[n, n+\delta]} x^2 dx \rightarrow \infty$ as $n \rightarrow \infty$.

Hahn decomposition: if ϕ is ASF, then there exists “positive set” P such that $\phi \geq 0$ on subsets of P and $\phi \leq 0$ on subsets of P^C .

Proof. P should maximize ϕ . Recall $\bar{V}(X) = \sup\{\phi(A) : A \in \Sigma\}$ for all k . Choose A_k with $\phi(A_k) > \bar{V}(X) - 2^{-k}$. Let $P = \liminf A_k$. Note $\underline{V}(A_k) = \bar{V}(A_k) - \phi(A_k)$. Fatou’s lemma gives $\underline{V}(A) \leq \liminf \underline{V}(A_k) = 0$. Also, $P^C = \limsup A_k^C$ and $\bar{V}(A_k^C) = \bar{V}(X) - \bar{V}(A_k) < 1/2^k$. Hence, $\bar{V}(P^C) = 0$ [$\sum \bar{V}(A_k^C) < \infty$] \square

Ex: if $\phi(A) = \int_A f$, then $P = \{f \geq 0\}$ or $P = \{f > 0\}$.

Compare to Jordan decomposition: $\phi = \bar{V} - \underline{V}$. Hahn makes this more precise. $\bar{V}(A) = \phi(A \cap P)$, $-\underline{V}(A) = \phi(A \setminus P)$ so $\underbrace{\bar{V}(A)}_{\bar{V}} - \underbrace{(-\phi(A \setminus P))}_{\underline{V}}$.

Lemma 6.4 (Level Decomposition of ASF). Suppose $\mu(X) < \infty$, $\phi \geq 0$ an ASF, then for all $a > 0$, there exists partition $X = Z \cup \bigcup_{k=1}^{\infty} E_k$, $\mu(Z) = 0$

$$a(k-1)\mu(A) \leq \phi(A) \leq ak\mu(A)$$

for all $A \subset E_k$

$$a(k-1) \leq \frac{\phi}{\mu} \leq ak$$

on E_k

Ex: if $\phi(A) = \int_A \frac{1}{\sqrt{x}} dx$ on $X = [0, 1]$ $a = 1$ $Z = \emptyset$, $E_1 = \emptyset$

Proof. WLOF, $a = 1$ (else consider ϕ/a). Apply Hahn to $\phi - k\mu$ ($k \in \mathbb{N}$) get a set P_k , $\phi \geq k\mu$ on P_k and $\phi \leq k\mu$ on P_k^C . Let $Q_k = \bigcup_{m \geq k} P_m$ note $\phi \geq k\mu$ on Q_k , $\phi \leq k\mu$ on Q_k . Let $Z = \bigcap_k Q_k$ on this set $\phi \geq k\mu$ so $\mu(Z) = 0$. Let $E_1 = Q_1^C$ then $\phi \leq \mu$ on E_1 , for $k > 1$, let $E_k = Q_{k-1} \setminus Q_k$, then $(k-1)\mu \leq \phi \leq k\mu$ as required. \square

Theorem 6.4 (Lebesgue Decomposition). *Let ϕ be an ASF and μ a σ -finite measure. Then there exists a unique additive set function α, σ such that $\phi = \alpha + \sigma$, α is AC wrt μ , σ is S wrt μ . Moreover, there exists $f \in L^1(X, \mu)$ such that $\alpha(A) = \int_A f$ and there exists Z such that $\mu(Z) = 0$ and $\sigma(A) = \phi(A \cap Z)$.*

Proof. Uniqueness: $\alpha + \sigma = \alpha' + \sigma'$, then $\alpha - \alpha' = \sigma' - \sigma$ so then both sides AC and S by equality, hence 0.

Existence of Z : comes from definition of singular. ASF $\mu(Z) = 0$ and $\phi = 0$ on Z^C . So $\sigma(A) = \sigma(A \cap Z)$. But $\alpha(A \cap Z) = 0$ because α is AC. Hence, $\sigma(A) = \phi(A \cap Z)$.

Hard parts existence of decomposition and existence of f .

Why do we care? Duality of functions spaces such as L^p . Dual space of L^p is $\{\psi : L^p \rightarrow \mathbb{C} : \psi \text{ linear cont.}\}$. ψ induces ASF $\phi(E) = \psi(\chi_E)$, $\phi(E \cup F) = \psi(\chi_E + \chi_F) = \phi(E) + \phi(F)$ if disjoint.

Why need to be σ -finite? ($X = \bigcup_{k \in \mathbb{N}} E_k$, $\mu(E_k) < \infty$)

Example 6.7. Let μ be counting measure on $[0, 1]$ (gives number of elements of set). $X = [0, 1]$. $\phi(A) = |A|$ is an ASF. If $\phi = \alpha + \sigma$, then $\sigma(A) = 0$ for all A (because the only Z with $\mu(Z) = 0$ is \emptyset). Hence, $\phi(A) = \int_A f d\mu$ for some $f \in L^1(\mu)$, which means $\sum_{x \in [0, 1]} |f(x)| < \infty$. But then $f \equiv 0$ except on a countable set C . But then get $\phi([0, 1] \setminus C) = 0$, contradicting $\phi(A) = |A|$.

Proof. (LDT) Can reduce to $\phi \geq 0$. For general case ϕ , $\phi = \bar{V} - \underline{V}$ so if $\bar{V} = \alpha_1 + \theta_1$, $\underline{V} = \alpha_2 + \theta_2$, then $\phi = (\alpha_1 - \alpha_2) + (\theta_1 - \theta_2)$.

2) Reduce to $\mu(X) < \infty$. Else $X = \bigcup_{\text{disjoint}} E_k$, If $\phi_k(A) = \phi(A)$ for $A \subset E_k$ has LD, $\alpha_k + \theta_k$, then let $\alpha(A) = \sum \alpha_k(A \cap E_k)$, $\sigma(A) = \sum \sigma_k(A \cap E_k)$.

3) Main case: $\phi \geq 0$, $\mu(X) < \infty$. Use level decomposition with $a = 2^{-m}$. Obtain $X = Z^m \cup \bigcup_{k=1}^{\infty} E_k^m$, $\mu(Z^m) = 0$ and $\phi(A)/\mu(A)$ is in $[(k-1)2^{-m}, k2^{-m}]$ for all $A \subset E_k^m$. Let $Z = \bigcup_{m=1}^{\infty} Z^m$, this will be our Z . Need $\phi(A \setminus Z) = \int_A f d\mu$. Let $f_m = \sum_k (k-1)2^{-m} \chi_{E_k^m}$. How close is $\int_A f_m$ to $\phi(A \setminus Z)$?

ϕ is AC wrt μ on each E_k^m , hence $\phi(A \cap E_k^m) = \phi((A \setminus Z) \cap E_k^m)$, null subsets of E^m are negligible. So $\phi(A) = \phi(A \cap Z) + \phi(A \setminus Z) = \phi(A \cap Z) + \sum_k \phi(A \cap E_k^m)$.

$\phi(A \cap E_k^m) \geq (k-1)2^{-m} \mu(A \cap E_k^m) = \int_{A \cap E_k^m} f_m$. $\phi(A \cap E_k^m) \leq k2^{-m} \mu(A \cap E_k^m) = \int_{A \cap E_k^m} f_m + 2^{-m} \mu(A \cap E_k^m)$. Sum over k

$$0 \leq \phi(A \setminus Z) - \int_A f_m d\mu \leq 2^{-m} \mu(X)$$

Remains to show f_m converge uniformly as $m \rightarrow \infty$ (then $\phi(A \setminus Z) = \int_A f d\mu$ with $f = \lim f_m$).

Observe E_k^m is covered by E_j^{m+1} for $j = 2k-2, 2k-1, 2k, 2k+1$ because

$$\left[\frac{k-1}{2^m}, \frac{k}{2^m} \right] \cap \left[\frac{j-1}{2^{m+1}}, \frac{j}{2^{m+1}} \right] = \emptyset$$

unless j is one of the values above. So if $f_m = (k-1)2^{-m}$ then $f_{m+1} = j2^{-m-1}$, where $2k-2 \leq j \leq 2k+1$. hence $|f_m - f_{m+1}| \leq 2^{-m}$. We get uniformly convergent of $\{f_m\}$. \square

Radon-Nikodym's Theorem:

Every AC ASF f wrt to σ -finite measure is of the form $\phi(A) = \int_A f d\mu$ for some $f \in L^1(\mu)$.

Lebesgue decomposition for measures: Let μ, ν be σ -finite measures on (X, Σ) . Then there exists measurable $f \geq 0$ and a measure σ such that σ is singular with respect to μ . $\nu(A) = \int_A f d\mu + \sigma(A)$.

Proof. Let $X = \bigcup E_j$, disjoint. $\nu(E_j) < \infty$. $\nu|_{E_j}$ is an ASF, so $\nu(A) = \int f_j d\mu + \sigma_j(A)$ for $A \subset E_j$, $f_j \in L^1(E_j, \mu)$. Note $f_j \geq 0$.

Let $\sigma(A) = \sum_j \sigma(A \cap E_j)$ and $f = \sum_j f_j \chi_{E_j}$ (gluing). Hence $\nu(A) = \sum_j \nu(A \cap E_j) = \sum_j (\int_{A \cap E_j} f_j d\mu + \sigma_j(A \cap E_j)) = \int_A f d\mu + \sigma(A)$. \square

6.4 The Dual Space of L^p

Definition. Let B a normed vector space over \mathbb{R} . Define

$$\begin{aligned} B^* &= \{ \text{linear } \psi : B \rightarrow \mathbb{R} : \exists c \text{ } |\psi(v)| \leq c \|v\| \forall v \in B \} \\ &= \{ \text{bounded linear functionals} \} \end{aligned}$$

is the dual space of L^p .

Theorem 6.5. Let (X, Σ, μ) be a σ -finite measure space, $1 \leq p < \infty$. Then there is a linear bijection between $L^p(X, \mu)^*$ and $L^{p'}(X, \mu)$ which preserves the norm.

Specifically, the bijection sends $g \in L^p$ to $\psi_g \in (L^p)^*$, $\psi_g(f) = \int_X fg d\mu$.

Example 6.8. $(X, \Sigma, \mu) = (\mathbb{N}, 2^{\mathbb{N}}, \text{counting})$, so $L^p(X, \mu)$ is ℓ^p : every $\psi : \ell^p \rightarrow \mathbb{R}$ is $\psi(x) = \sum_{k=1}^{\infty} x_k y_k$ for some $y \in \ell^{p'}$. Note so for ℓ^∞ . One can define $\psi(x) = \lim_{k \rightarrow \infty} x_k$ when exists, and extend to a bounded linear functional on ℓ^∞ . Suppose we had $y \in \ell^1$ such that $\sum_1^{\infty} x_k y_k = \lim_{k \rightarrow \infty} x_k$ for all convergent sequences x . Problem is limit does not depend on any particular x_k . So can choose one and add one to it, etc. so we must have $y_k = 0$. But this is true for all of them so that we must have $y = 0$ but then if $x = 1$ for all k so sum is not equal to the limit, which is 1, a contradiction.

Proof. If $g \in L^{p'}$, then $|\int fg| \leq \|f\|_p \|g\|_{p'}$, so ψ_g is bounded: $|\psi_g(f)| \leq \|g\|_{p'} \|f\|_p$. So $\|\psi_g\| \leq \|g\|_{p'}$. This gives a linear map $L^{p'} \rightarrow (L^p)^*$. This map is injective: if $\int fg = 0$ for all $f \in L^p$, then $g \equiv 0$ a.e., LTR. Surjectivity: σ -finite case can be reduced to $\mu(X) < \infty$ by $X = \bigcup_1^{\infty} E_j$, $E_1 \subset E_2 \subset E_3 \subset \dots$.

We get g_j on E_j from the finite case ($\mu(E_j) < \infty$). Glue them into one g which gives $\psi(f) = \int_X fg$. Finite case: $\mu(X) < \infty$. Define an ASF $\phi(A) = \psi(\chi_A)$. It is AC wrt

$\mu : \mu(A) = 0$ measure $\chi_A = 0$ a.e., so $\psi(\chi_A) = 0$. By Radon-Nikodym, there exists $g \in L^1(X, \mu)$ such that $\phi(A) = \int_A g \, d\mu$ for all $A \subset X$. This means $\psi(\chi_A) = \int g \chi_A \, d\mu$. $\psi(f) = \int g f \, d\mu$ for all simple f . Once we know that $g \in L^p$, we can conclude the proof as follows: ψ and ψg are continuous on L^p and $\psi = \psi g$ on a dense subset (simple functions), hence $\psi = \psi g$.

case $p = 1, p' = \infty$. Case $1 < p < \infty, 1 < p' < \infty$ (textbook). for the first then, $\phi(A) = \psi(\chi_A) \leq \|\psi\| \cdot \|\chi_A\|_1 = \|\psi\| \mu(A)$. Hence, $\phi(A)/\mu(A) \leq \|\psi\|$. In the proof of Lebesgue decomposition, we will not have density $> \|\psi\|$ because of the inequality just shown. Thus $\|g\|_{p'} \leq \|\psi\|$. \square

7 Quiz Problems

Quiz 1. Suppose A is a closed proper subset of $[0, 1]$ (so $A \subset [0, 1]$ and $A \neq [0, 1]$). Is $|A|_e < 1$?

Quiz 2. Suppose $A \subset \mathbb{R}$ is a set such that $A \cap [-k, k]$ is measurable for every $k \in \mathbb{N}$. Is it the case that A is measurable?

Quiz 3. Suppose $A_1 \supset A_2 \supset A_3 \supset \cdots$ is a nested sequence of measurable subsets of \mathbb{R}^n such that $|A_1| < \infty$. Is it the case that $|A_k \setminus A_{k+1}| \rightarrow 0$ as $k \rightarrow \infty$?

Quiz 4. Suppose $E \subset \mathbb{R}^n$ is a set such that $|A \cap E|_e + |A \cap E^c|_e \leq |A|_e$ for every $A \subset \mathbb{R}^n$. Is it the case that E is measurable?

Quiz 5. Let $f(x)$ be the function

$$f(x) = \begin{cases} 1, & x \in \mathbb{Q} \\ 0, & \text{otherwise} \end{cases}$$

Is it the case that f is a measurable function on \mathbb{R} ?

Quiz 6. Suppose $f : \mathbb{R} \rightarrow \mathbb{R}$ is measurable. Define

$$g(x) = \begin{cases} f(x), & |f(x)| \geq 1 \\ 0, & \text{otherwise} \end{cases}$$

Is g necessarily a measurable function on \mathbb{R} ?

Quiz 7. Let $f(x)$ be the function given by

$$f(x) = \begin{cases} \sin(1/x), & x \neq 0 \\ 0, & x = 0 \end{cases}$$

Is f upper semicontinuous on \mathbb{R} ?

Quiz 8. Suppose that $\{f_k : [0, 1] \rightarrow \mathbb{R}\}$ is a collection of simple functions which converge uniformly to f . Is f necessarily a simple function?

Quiz 9. Suppose that $f : [0, \infty) \rightarrow [0, \infty)$ is measurable. Is it the case that

$$\int_{[0, \infty)} e^{-x/k} f(x) \, dx \longrightarrow \int_{[0, \infty)} f(x) \, dx$$

as k tends to infinity?

Quiz 10. Suppose that $f : [0, \infty) \rightarrow [0, \infty)$ is measurable and $\int_{[0, \infty)} f < \infty$. Is it true that $\int_{[k, \infty)} f(x) \, dx \rightarrow 0$ as $k \rightarrow \infty$?

8 Quiz Solutions

Solution Quiz 1. Suppose that $f : [0, \infty) \rightarrow [0, \infty)$ is measurable and $\int_{[0, \infty)} f < \infty$. Is it true that $\int_{[k, \infty)} f(x) dx \rightarrow 0$ as $k \rightarrow \infty$?

Proof. Let $f_k := f\chi_{[k, \infty)}$. Then goes to 0 as $k \rightarrow \infty$ and f dominates f_k and $f_k \rightarrow 0$.

Another idea: $f\chi_{(0, k)} \nearrow f$ by MCT, $\int \chi_{(0, k)} f \rightarrow \int f$.

Or

$R(f, E)$ has finite measure, have nested subsets from k to infinite areas, nested, contained, measure sets tend to measure of intersection which is 0?

Solution Quiz 2. Suppose that $f : \mathbb{R} \rightarrow \mathbb{R}$ is measurable and the function $g(x, y) = f(x)f(y)^2$ is integrable on \mathbb{R}^2 . Is f integrable on \mathbb{R}^2 , i.e. is $f \in L^1(\mathbb{R})$?

Proof. We do indeed have $f \in L^1(\mathbb{R})$. The function given by $x \mapsto f(x)f(y)^2$ is integrable a.e. on y . There are two cases: $f(y) = 0$ for almost all y or there exists y such that f is integrable and $f(y) \neq 0$. In either case, f is integrable. \square

Solution Quiz 3. Let $f(x) = 2$ and $g(x) = 3$ for all $x \in \mathbb{R}$. Is the convolution $f * g$ equal to 6 everywhere on \mathbb{R} ?

Proof. It is the case that $f * g = 6$ as

$$(f * g)(x) = \int_{\mathbb{R}} f(x-t)g(t) dt = \int_{\mathbb{R}} 6 dt = \infty.$$

Solution Quiz 4. Suppose that $f, g \in L^\infty(\mathbb{R}^n)$. Is it the case that $fg \in L^\infty(\mathbb{R}^n)$?

Proof. We have $|fg| = |f||g| \leq |f|\|g\|_\infty$ for almost all $x \in \mathbb{R}^n$. But then

$$\int |fg| \leq \int |f|\|g\|_\infty = \|g\|_\infty \int |f| = \|f\|_\infty \|g\|_\infty.$$

Alternatively, $f, g \in L^\infty(\mathbb{R}^n)$ implies that f, g are bounded almost everywhere. Suppose $|f| \leq M$ and $|g| \leq N$ for almost all $x \in \mathbb{R}^n$. But then $|fg| \leq MN$ for almost all $x \in \mathbb{R}^n$. Therefore, $fg \in L^\infty(\mathbb{R}^n)$. Note that it is not generally true that if $f, g \in L^p$, for some $p < \infty$, that $fg \in L^p$. For example, $\frac{1}{x^r} \in L^p((0, 1))$ if and only if $pr < 1$ if and only if $p < 1/r$. Then generally, $(1/x^r)^2$ is not in the same L^p . \square

Solution Quiz 5. Suppose that $f, g : \mathbb{R} \rightarrow \mathbb{C}$ are measurable functions such that $|f|^4$ and $|g|^{4/3}$ are integrable. Prove or disprove: fg is integrable.

Proof. We know $\int |fg| \leq \|f\|_4 \|g\|_{4/3}$ and $\frac{1}{4} + \frac{3}{4} = 1$. This fails if f, g are not measurable. Take $g = 1$ and f a $(-1)^n$ type of deal on a nonmeasurable set. \square

Solution Quiz 6. Suppose that $0 < p \leq \infty$ and that the sequence $\{a_k\}$ belongs to ℓ^p . Is it the case that every subsequence $\{a_{k_j}\}$ belongs to ℓ^p ?

Proof. $p < \infty$, $\sum |a_{k_j}|^p \leq \sum |a_k|^p$ and $p = \infty$, $a_k \leq M$ then $a_{k_j} \leq M$.

Solution Quiz 7. For $k \in \mathbb{N}$, let $f_k = \sqrt{k}\chi_{[0,1/k]}$. Is it the case that $f_k \rightarrow 0$ in $L^2(\mathbb{R})$?

Proof. This is not the case as

$$\|\sqrt{k}\chi_{[0,1/k]}\|_2 = \left(\int k\chi_{[0,1/k]} \right)^{1/2} = 1.$$

However, this does converge for $p < 2$ as

$$\int_0^{1/k} k^{p/2} = k^{p/2-1} \rightarrow 0$$

if $p < 2$. \square

9 Homework Problems

Problem 1. Given an arbitrary set $A \subset \mathbb{R}$ and a number $c > 0$, let $B = \{ca : a \in A\}$. Prove that $|B|_e = c |A|_e$.

Problem 2. Suppose that a set $A \subset \mathbb{R}$ is measurable. Prove that for every $c > 0$ the set $B = \{ca : a \in A\}$ is also measurable.

Problem 3. Given a sequence of continuous functions $f_k : \mathbb{R} \rightarrow \mathbb{R}$, let B be the set of all points $x \in \mathbb{R}$ such that the sequence $\{f_k(x)\}$ is bounded. Prove that B is a measurable set. [Hint: try to construct B from the sets $\{x : |f_k(x)| \leq M\}$ by using countable unions and intersections.]

Problem 4. Given a sequence of continuous functions $f_k : \mathbb{R} \rightarrow \mathbb{R}$, let C be the set of all points $x \in \mathbb{R}$ such that $\lim_{k \rightarrow \infty} f_k(x) = 0$. Prove that C is a measurable set.

Problem 5. Prove that the set

$$A = \{x \in \mathbb{R} : \exists k \in \mathbb{N} \text{ such that } |2^x - 2^k| \leq 1\}$$

is measurable and $|A| < \infty$.

Problem 6. Prove that the set

$$A = \{x \in [0, 1] : \forall q \in \mathbb{N} \exists p \in \mathbb{N} \text{ such that } |x - p/q| \leq 1/q^2\}$$

is measurable and $|A| = 0$.

Problem 7. Suppose E and Z are sets in \mathbb{R}^n such that $E \cup Z$ is measurable and $|Z| = 0$. Prove that E is measurable.

Problem 8. Given a continuous function $f : \mathbb{R}^n \rightarrow \mathbb{R}^n$, define $\mathcal{M} = \{E \subset \mathbb{R}^n : f^{-1}(E) \text{ is Borel}\}$.

(a) Prove that \mathcal{M} is a σ -algebra.

(b) Prove that if E is Borel, then $f^{-1}(E)$ is Borel. [Hint: Use (a).]

Problem 9. Suppose $f: \mathbb{R} \rightarrow \mathbb{R}$ is a function with a continuous derivative. Prove that for every measurable set E , the set $f(E)$ is also measurable. [Hint: although f need not be Lipschitz, its restriction to any bounded interval is Lipschitz.]

Problem 10. Given a set $E \subset [0, \infty)$, define a function $f: [0, \infty) \rightarrow [0, \infty)$ by $f(x) = |E \cap [0, x]|_e$.

- (a) Prove that f is Lipschitz continuous.
- (b) Prove that for every number b with $0 < b < |E|_e$ there exists a set $F \subset E$ such that $|F|_e = b$.

Problem 11. Show that there exists a nested sequence of sets $E_1 \supset E_2 \supset \cdots$ such that $|E_1|_e < \infty$ and $\bigcap_{k=1}^{\infty} E_k = \emptyset$ but $\lim_{k \rightarrow \infty} |E_k|_e > 0$; that is, outer measure is not continuous under nested intersections. [Hint: Use the translates of the Vitali set.]

Problem 12. Show that for the standard middle-third Cantor set $C \subset [0, 1]$, the difference set $C - C$ contains a neighborhood of 0. [Hint: C is the intersection of nested sets C_n where $C_0 = [0, 1]$ and $C_{n+1} = \frac{1}{3}C_n \cup (\frac{1}{3}C_n + \frac{2}{3})$. Find $C_n - C_n$ using induction.]

Remark: This shows that having $|E| > 0$ is not necessary for $E - E$ to contain a neighborhood of 0.

Problem 13. Suppose that $f: \mathbb{R}^n \rightarrow \mathbb{R}$ is a function such that $f(\mathbb{R}^n)$ is countable and $f^{-1}(t)$ is measurable for every $t \in \mathbb{R}$. Prove that f is measurable.

Problem 14. Prove that without the assumption “ $f(\mathbb{R}^n)$ is countable” the statement in Problem 13 would not be true.

Problem 15. Suppose that $f: \mathbb{R} \rightarrow \mathbb{R}$ is measurable, and $g: \mathbb{R} \rightarrow \mathbb{R}$ is continuously differentiable with $g' > 0$ everywhere. Prove that $f \circ g$ is measurable.

Problem 16.

- (a) Suppose $f: \mathbb{R}^n \rightarrow \mathbb{R}$ is a continuous function such that f^2 is measurable. Prove that f is measurable.
- (b) Prove that the statement in (a) is false if f is not assumed continuous.

Problem 17. Suppose that $f : E \rightarrow \mathbb{R}$ is a measurable function, where $E \subset \mathbb{R}^n$ is measurable.

- (a) Prove that there exists a Borel set $H \subset E$ such that the restriction $f|_H$ is Borel measurable and $|E \setminus H| = 0$.
- (b) If, in addition, E is a Borel set, prove that there exists a Borel measurable function $g : E \rightarrow \mathbb{R}$ such that $f = g$ a.e.. [Hint: For part (a), take a countable union of closed sets obtained from Lusin's Theorem.]

Problem 18. Suppose $\phi : [0, \infty) \rightarrow [0, \infty)$ is a function such that $\phi(t) \rightarrow 0$ as $t \rightarrow \infty$. Consider a sequence of measurable functions $\{f_k\}$, $f_k : \mathbb{R}^n \rightarrow \mathbb{R}$, such that $|f_k(x)| \leq \phi(|x|)$ for every k , and $f_k \rightarrow f$ a.e.. Prove that the conclusion of Egorov's Theorem holds in this situation; that is, for every $\epsilon > 0$, there exists a closed set $E(\epsilon) \subset \mathbb{R}^n$ such that $|\mathbb{R}^n \setminus E(\epsilon)| < \epsilon$ and $f_k \rightarrow f$ uniformly on $E(\epsilon)$. [Hint: Follow the proof of Egorov's theorem.]

Problem 19. Suppose that $f : \mathbb{R}^n \rightarrow \mathbb{R}$ is a function such that $f(\mathbb{R}^n)$ is countable and $f^{-1}(t)$ is measurable for every $t \in \mathbb{R}$. Prove that f is measurable.

Problem 20. Prove that without the assumption " $f(\mathbb{R}^n)$ is countable" the statement in Problem 19 would not be true.

10 Homework Solutions

Solution Problem 1. Given an arbitrary set $A \subset \mathbb{R}$ and a number $c > 0$, let $B = \{ca : a \in A\}$. Prove that $|B|_e = c |A|_e$.

Proof. Given $\epsilon > 0$, let $\{[s_k, t_k]\}$ be a countable cover of A such that

$$\sum_k (t_k - s_k) \leq |A|_e + \epsilon$$

The intervals $[cs_k, ct_k]$ cover B , since every point of B is of the form ca where $a \in A$ is covered by some interval $[s_k, t_k]$. Therefore,

$$|B|_e \leq \sum_k (ct_k - cs_k) = c \sum_k (t_k - s_k) \leq c |A|_e + \epsilon$$

Since $\epsilon > 0$ was arbitrary, it follows that $|B|_e \leq c |A|_e$.

It remains to observe that $A = c^{-1}B$, which by the above implies $|A|_e \leq c^{-1} |B|_e$, i.e., $|B|_e \geq c |A|_e$. Thus, $|B|_e = c |A|_e$. \square

Solution Problem 2. Suppose that a set $A \subset \mathbb{R}$ is measurable. Prove that for every $c > 0$ the set $B = \{ca : a \in A\}$ is also measurable.

Proof. Given $\epsilon > 0$, let G be an open set that contains A and satisfies $|G \setminus A|_e < \epsilon/c$. Since the function $f(x) = x/c$ is continuous, the preimage of G under this function is also open. This preimage $f^{-1}(G)$ is cG . Since $A \subset G$, it follows that $B \subset cG$. Moreover by Problem 1,

$$|(cG) \setminus B|_e = |c(G \setminus A)|_e = c |G \setminus A|_e < \epsilon.$$

Since ϵ was arbitrary, this proves that B is measurable. \square

Solution Problem 3. Given a sequence of continuous functions $f_k : \mathbb{R} \rightarrow \mathbb{R}$, let B be the set of all points $x \in \mathbb{R}$ such that the sequence $\{f_k(x)\}$ is bounded. Prove that B is a measurable set. [Hint: try to construct B from the sets $\{x : |f_k(x)| \leq M\}$ by using countable unions and intersections.]

Proof. For $k, m \in \mathbb{N}$, let

$$A(k, m) = \{x \in \mathbb{R} : |f_k(x)| \leq m\} = f_k^{-1}([-m, m]).$$

Being the preimage of a closed set under a continuous function, $A(k, m)$ is closed, and in particular measurable. Let

$$A := \bigcup_{m=1}^{\infty} \bigcap_{k=1}^{\infty} A(k, m),$$

which is also measurable, being obtained from measurable sets by countable set operations. We claim that $A = B$.

If $x \in A$, then there exists $m \in \mathbb{N}$ such that $|f_k(x)| \leq m$ for all $k \in \mathbb{N}$, which shows the sequence $\{f_k(x)\}$ is bounded.

Conversely, if the sequence $\{f_k(x)\}$ is bounded, then there exists $m \in \mathbb{N}$ such that all elements of the sequence are at most m in absolute value. This means $|f_k(x)| \leq m$ for all k , hence $x \in A$. \square

Solution Problem 4. Given a sequence of continuous functions $f_k: \mathbb{R} \rightarrow \mathbb{R}$, let C be the set of all points $x \in \mathbb{R}$ such that $\lim_{k \rightarrow \infty} f_k(x) = 0$. Prove that C is a measurable set.

Proof. For $k, m \in \mathbb{N}$, let

$$A(k, m) = \{x \in \mathbb{R} : |f_k(x)| < 1/m\} = f_k^{-1}((-1/m, 1/m)).$$

Being the preimage of an open set under a continuous function, $A(k, m)$ is open, and in particular measurable. Let

$$A := \bigcap_{m=1}^{\infty} \bigcup_{N=1}^{\infty} \bigcap_{k=N}^{\infty} A(k, m).$$

This set is also measurable, being obtained from measurable sets by countable set operations. We claim that $A = C$.

Suppose $x \in A$. Given $\epsilon > 0$, pick $m \in \mathbb{N}$ such that $1/m \leq \epsilon$. Since $x \in \bigcup_{N=1}^{\infty} \bigcap_{k=N}^{\infty} A(k, m)$, there exists N such that $x \in \bigcap_{k=N}^{\infty} A(k, m)$, which means $|f_k(x)| < 1/m$ for all $k \geq N$. Thus, $|f_k(x)| < \epsilon$ for all $k \geq N$, which proves $\lim_{k \rightarrow \infty} f_k(x) = 0$.

Conversely, suppose $x \in C$. Given $m \in \mathbb{N}$, use the definition of the limit $\lim_{k \rightarrow \infty} f_k(x) = 0$ to find N such that $|f_k(x)| < 1/m$ for all $k \geq N$. The latter means $x \in \bigcap_{k=N}^{\infty} A(k, m)$. Therefore, for every $m \in \mathbb{N}$ the inclusion $x \in \bigcup_{N=1}^{\infty} \bigcap_{k=N}^{\infty} A(k, m)$ holds. This means $x \in A$. \square

Solution Problem 5. Prove that the set

$$A = \{x \in \mathbb{R} : \exists k \in \mathbb{N} \text{ such that } |2^x - 2^k| \leq 1\}$$

is measurable and $|A| < \infty$.

Proof. For each $k \in \mathbb{N}$, the inequality $|2^x - 2^k| \leq 1$ is equivalent to $\log_2(2^k - 1) \leq x \leq \log_2(2^k + 1)$. Thus, $A = \bigcup_{k=1}^{\infty} I_k$, where $I_k = [\log_2(2^k - 1), \log_2(2^k + 1)]$. Begin an interval, each I_k is measurable. Hence, A is measurable. By countable subadditivity of measure, $|A| \leq \sum_{k=1}^{\infty} |I_k|$. It remains to show the series $\sum_{k=1}^{\infty} |I_k|$ converges. This can be done by the

comparison test, limit comparison test, or the ratio test. We use the Limit Comparison Test with $\sum_{k=1}^{\infty} 2^{-k}$ as a reference series:

$$\begin{aligned} \frac{|I_k|}{2^{-k}} &= \frac{\log_2(2^k + 1) - \log_2(2^k - 1)}{2^{-k}} \\ &= \frac{k + \log_2(1 + 2^{-k}) - (k + \log_2(1 - 2^{-k}))}{2^{-k}} \\ &= \frac{\log_2(1 + 2^{-k}) - \log_2(1 - 2^{-k})}{2^{-k}} \\ &= \frac{1}{\log 2} \left\{ \frac{\log(1 + 2^{-k})}{2^{-k}} + \frac{\log(1 - 2^{-k})}{-2^{-k}} \right\} \xrightarrow{k \rightarrow \infty} \frac{2}{\log 2} \end{aligned}$$

Here the last step is based on $\lim_{x \rightarrow 0} \frac{\log(1+x)}{x} = 1$. Since $\sum_{k=1}^{\infty} 2^{-k}$ converges, so does $\sum_{k=1}^{\infty} |I_k|$. \square

Solution Problem 6. Prove that the set

$$A = \{x \in [0, 1] : \forall q \in \mathbb{N} \exists p \in \mathbb{N} \text{ such that } |x - p/q| \leq 1/q^2\}$$

is measurable and $|A| = 0$.

Proof. Let $A_q = \bigcup_{p=1}^{\infty} E(p, q)$, where $E(p, q) = \left[\frac{p}{q} - \frac{1}{q^2}, \frac{p}{q} + \frac{1}{q^2} \right] \cap [0, 1]$. This is a countable union of measurable sets $E(p, q)$ (which are intervals, possibly empty), so it is measurable. Then the set $A = \bigcap_{q \in \mathbb{N}} A_q$ is measurable too.

We have $|E(p, q)| \leq 2/q^2$ by construction of $E(p, q)$. Also, when $p > q + 1$, we have $\frac{p}{q} - \frac{1}{q^2} \geq 1 + \frac{1}{q} - \frac{1}{q^2} \geq 1$, which implies $E(p, q) = \emptyset$. By subadditivity,

$$|A_q| \leq \sum_{p=1}^{\infty} |E(p, q)| \leq \sum_{p=1}^{q+1} \frac{2}{q^2} = \frac{2q+2}{q^2}$$

By monotonicity, $|A| \leq |A_q|$ for each q . Since $|A_q| \xrightarrow{q \rightarrow \infty} 0$, it follows that $|A| = 0$. \square

Solution Problem 7. Suppose E and Z are sets in \mathbb{R}^n such that $E \cup Z$ is measurable and $|Z| = 0$. Prove that E is measurable.

Proof. Since $Z \setminus E \subset Z$, the monotonicity of outer measure implies $|Z \setminus E|_e = 0$, hence $Z \setminus E$ is measurable. Then

$$E = (E \cup Z) \setminus (Z \setminus E)$$

is measurable, being the difference of two measurable sets. [This could be done with Carathéodory's Theorem or with the " G_δ minus a null set" theorem, but it is easier without.] \square

Solution Problem 8. Given a continuous function $f: \mathbb{R}^n \rightarrow \mathbb{R}^n$, define $\mathcal{M} = \{E \subset \mathbb{R}^n : f^{-1}(E) \text{ is Borel}\}$.

- (a) Prove that \mathcal{M} is a σ -algebra.
- (b) Prove that if E is Borel, then $f^{-1}(E)$ is Borel. [Hint: Use (a).]

Proof.

- (a) This proof does not involve f being continuous; the argument works for any map f . Taking preimages commutes with any set operations: for example,

$$f^{-1}(E^C) = \{x : f(x) \in E^C\} = \{x : f(x) \notin E\} = (f^{-1}(E))^C$$

and

$$f^{-1}\left(\bigcup_i E_i\right) = \{x : \exists i f(x) \in E_i\} = \bigcup_i f^{-1}(E_i).$$

If $E \in \mathcal{M}$, then $f^{-1}(E^C) = f^{-1}(E)^C$ is the complement of a Borel set, and hence is Borel. Hence, $E^C \in \mathcal{M}$. Also, if $E_k \in \mathcal{M}$ for each $k \in \mathbb{N}$, then

$$f^{-1}\left(\bigcup_k E_k\right) = \bigcup_k f^{-1}(E_k)$$

is the countable union of Borel sets, and hence is Borel. Therefore, $\bigcup_k E_k \in \mathcal{M}$.

The definition of a σ -algebra in the book also requires us to check that \mathcal{M} is nonempty: to do this, it suffices to notice that $f^{-1}(\emptyset) = \emptyset$ is Borel, hence $\emptyset \in \mathcal{M}$.

- (b) Since f is continuous, the preimage of any open set under f is open, and is hence Borel. This means \mathcal{M} contains all open sets. By definition, the Borel σ -algebra \mathcal{B} is the *smallest* σ -algebra that contains all open sets. Thus, $\mathcal{B} \subset \mathcal{M}$, which by definition of \mathcal{M} means that $f^{-1}(E)$ is Borel whenever E is Borel.

[It is tempting to approach statement (b) by "writing a Borel set E in terms of open/closed sets" and concluding that $f^{-1}(E)$ can also be written in this way. But there is no such structural formula for Borel sets: one can only get the proper subclasses like G_δ , $G_{\delta\sigma}$, $G_{\delta\sigma\delta}$, and so on. The whole story is complicated: see [Borel hierarchy](#) on Wikipedia]. \square

Solution Problem 9. Suppose $f: \mathbb{R} \rightarrow \mathbb{R}$ is a function with a continuous derivative. Prove that for every measurable set E , the set $f(E)$ is also measurable. [Hint: although f need not be Lipschitz, its restriction to any bounded interval is Lipschitz.]

Proof. For each $j \in \mathbb{N}$, the set $E_j = E \cap [-j, j]$ is measurable, as the intersection of two measurable sets is measurable. Since $E = \bigcup_j E_j$, it follows that $f(E) = \bigcup_j f(E_j)$. So it suffices to prove $f(E_j)$ is measurable for every j .

The derivative f' , being continuous, is bounded on the interval $[-j, j]$. By the Mean Value Theorem, f is Lipschitz on $[-j, j]$: indeed, $|f(a) - f(b)| \leq |a - b| \sup_{[-j, j]} |f'|$. A technical detail arises: we only proved the measurability of images for Lipschitz functions on all of \mathbb{R}^n . To get around this, define

$$f_j(x) = \begin{cases} f(x), & x \in [-j, j] \\ f(-j), & x < -j \\ f(j), & x > j \end{cases}$$

Such an extended function f_j is Lipschitz continuous on all of \mathbb{R} . Indeed in each of the three closed intervals $(-\infty, -j]$, $[-j, j]$, $[j, \infty)$, the Lipschitz condition holds by construction. For arbitrary $a < b$, partition the interval $[a, b]$ by the points $\{-j, j\}$ should they lie there, and apply the Lipschitz continuity to each interval, and use the triangle inequality.

Conclusion: $f_j(E_j)$, which is the same as $f(E_j)$, is measurable. \square

[Note: In fact, for every set $E \subset \mathbb{R}^n$, any Lipschitz function $f: E \rightarrow \mathbb{R}^n$ can be extended to a Lipschitz function $F: \mathbb{R}^n \rightarrow \mathbb{R}^n$. Therefore, when discussing the measurability of $f(E)$, it suffices to check that f is Lipschitz on the set E .

Proof (Sketch). It suffices to extend a real-valued Lipschitz function $f: E \rightarrow \mathbb{R}$, because the vector-valued case follows by extending each component. Let L be the Lipschitz constant of f , and define, for every $x \in \mathbb{R}^n$,

$$F(x) = \inf_{a \in E} (f(a) + L|x - a|).$$

It is an exercise with the definition of \inf to prove that F is Lipschitz with constant L , and that $F(x) = f(x)$ when $x \in E$. \square

Remark: Extending a map $f: E \rightarrow \mathbb{R}^n$ in the above fashion, one finds the Lipschitz constant of the extension is at most $L\sqrt{n}$, where L is the Lipschitz constant of the original map. There is a deeper extension theorem (due to Kirszbraun) according to which an extension with the same Lipschitz constant L exists.]

Solution Problem 10. Given a set $E \subset [0, \infty)$, define a function $f: [0, \infty) \rightarrow [0, \infty)$ by $f(x) = |E \cap [0, x]|_e$.

- (a) Prove that f is Lipschitz continuous.
- (b) Prove that for every number b with $0 < b < |E|_e$ there exists a set $F \subset E$ such that $|F|_e = b$.

Proof.

- (a) We claim that $0 \leq f(b) - f(a) \leq b - a$ for any $a, b \in [0, \infty)$ such that $a < b$; this yields the Lipschitz continuity with constant 1. On one hand, $f(b) \geq f(a)$ by the monotonicity of outer measure: $E \cap [0, a] \subset E \cap [0, b]$. On the other, $E \cap [0, b] \subset (E \cap [0, a]) \cup [a, b]$, which implies

$$f(b) \leq |(E \cap [0, a]) \cup [a, b]|_e \leq |E \cap [0, a]|_e + |[a, b]| = f(a) + (b - a)$$

by subadditivity.

- (b) By Theorem 3.27 in the textbook, the outer measure is continuous under nested unions even if the sets are not measurable. Since $E = \bigcup_{k \in \mathbb{N}} (E \cap [0, k])$, it follows that

$$|E|_e = \lim_{k \rightarrow \infty} |E \cap [0, k]|_e = \lim_{k \rightarrow \infty} f(k).$$

Since $b < |E|_e$, by the definition of limit there exists k such that $f(k) > b$. Also, $f(0) = |E \cap \{0\}| = 0$. Applying the Intermediate Value Theorem to f on the interval $[0, k]$ (which is possible since f is continuous by part (a)), we conclude that there exists $x \in (0, k)$ such that $f(x) = b$. Then the set $F = |E \cap [0, x]|$ meets the requirements.

□

Solution Problem 11. Show that there exists a nested sequence of sets $E_1 \supset E_2 \supset \cdots$ such that $|E_1|_e < \infty$ and $\bigcap_{k=1}^{\infty} E_k = \emptyset$ but $\lim_{k \rightarrow \infty} |E_k|_e > 0$; that is, outer measure is not continuous under nested intersections. [Hint: Use the translates of the Vitali set.]

Proof. Let $V \subset [0, 1]$ be the Vitali set described in these notes: recall that $|V|_e > 0$ and that the sets $V + q$ are disjoint for all $q \in \mathbb{Q}$. Let

$$E_k = \bigcup_{j=k}^{\infty} \left(V + \frac{1}{j} \right).$$

Then $E_1 \subset V + [0, 1] \subset [0, 2]$, hence $|E_1|_e \leq 2 < \infty$.

Suppose $x \in \bigcap_{k=1}^{\infty} E_k$. This means that for each $k \in \mathbb{N}$ there exists $j \geq k$ such that $x \in V + 1/j$. In particular, $x \in V + 1/j$ for infinitely many distinct values of j . But this is impossible as the sets $V + 1/j$ are disjoint, a contradiction. Therefore, $\bigcap_{k=1}^{\infty} E_k$ is empty.

The sets E_k are nested by construction, hence $|E_k|_e$ is a non-increasing sequence. It is bounded from below by $|V|_e$ because each E_k contains a translated copy of V . Thus, $\lim_{k \rightarrow \infty} |E_k|_e \geq |V|_e > 0$. \square

Solution Problem 12. Show that for the standard middle-third Cantor set $C \subset [0, 1]$, the difference set $C - C$ contains a neighborhood of 0. [Hint: C is the intersection of nested sets C_n where $C_0 = [0, 1]$ and $C_{n+1} = \frac{1}{3}C_n \cup (\frac{1}{3}C_n + \frac{2}{3})$. Find $C_n - C_n$ using induction.]

Remark: This shows that having $|E| > 0$ is not necessary for $E - E$ to contain a neighborhood of 0.

Proof. Recall that $A - B = \{a - b : a \in A, b \in B\}$. This definition implies that

$$(A_1 \cup A_2) - (B_1 \cup B_2) = \bigcup_{i,j=1}^2 (A_i - B_j) \quad (*)$$

Furthermore, for any $t \in \mathbb{R}$ we have $(A + t) - B = (A + B) + t$, $A - (B + t) = (A - B) - t$, and $tA - tB = t(A - B)$; all these follow directly from the definition.

The equality $C_0 - C_0 = [-1, 1]$ holds because, on one hand, $|x - y| \leq 1$ when $x, y \in [0, 1]$, while on the other, $C_0 - C_0 \supset [0, 1] - \{0, 1\} = [0, 1] \cup [-1, 0] = [-1, 1]$.

Assume $C_n - C_n = [-1, 1]$. Use the relation $C_{n+1} = \frac{1}{3}C_n \cup (\frac{1}{3}C_n + \frac{2}{3})$ and distribute the difference according to (*) and other properties stated at the beginning:

$$\begin{aligned} C_{n+1} - C_{n+1} &= \left(\frac{1}{3}C_n - \frac{1}{3}C_n\right) \cup \left(\frac{1}{3}C_n - \frac{1}{3}C_n + \frac{2}{3}\right) \cup \left(\frac{1}{3}C_n - \frac{1}{3}C_n - \frac{2}{3}\right) \\ &= [-1/3, 1/3] \cup ([-1/3, 1/3] + 2/3) \cup ([-1/3, 1/3] - 2/3) \\ &= [-1/3, 1/3] \cup [1/3, 1] \cup [-1, -1/3] = [-1, 1] \end{aligned}$$

The set $(\frac{1}{3}C_n + \frac{2}{3}) - (\frac{1}{3}C_n + \frac{2}{3})$ is not included above because it is the same as $(\frac{1}{3}C_n - \frac{1}{3}C_n)$. By induction, $C_n - C_n = [-1, 1]$ for all n .

Since $C \subset C_n$ for every n , it follows that $C - C \subset [-1, 1]$. To prove the reverse inclusion, fix $a \in [-1, 1]$. For each n , there exist $x_n, y_n \in C_n$ such that $x_n - y_n = a$. Since all these numbers are contained in $[0, 1]$, we can pick a convergent subsequence $\{x_{n_k}\}$. So, $x_{n_k} \rightarrow x$ and since $x_{n_k} - y_{n_k} = a$, we also have $y_{n_k} \rightarrow y$ where y is such that $x - y = a$.

It remains to prove that $x, y \in C$. For each $m \in \mathbb{N}$ we have $x_{n_k}, y_{n_k} \in C_m$ for $k \geq m$ by construction. Since C_m is compact, it follows that $x, y \in C_m$. And since this holds for every $m \in \mathbb{N}$, we have $x, y \in C$. \square

Solution Problem 13. Suppose that $f : \mathbb{R}^n \rightarrow \mathbb{R}$ is a function such that $f(\mathbb{R}^n)$ is countable and $f^{-1}(t)$ is measurable for every $t \in \mathbb{R}$. Prove that f is measurable.

Proof. Let $B = f(\mathbb{R}^n)$, a countable subset of \mathbb{R} . For any $a \in \mathbb{R}$ we have

$$\{f > a\} = \bigcup_{b \in B, b > a} f^{-1}(b),$$

which is a countable union of measurable sets, and is hence measurable. The domain of f , which is \mathbb{R}^n , is also measurable. Thus f is measurable. \square

Solution Problem 14. Prove that without the assumption “ $f(\mathbb{R}^n)$ is countable” the statement in Problem 13 would not be true.

Proof. The statement in Problem 13 is made for any n . To disprove it, it suffices to show it fails for some n . With massive loss of generality, let $n = 1$. Let $V \subset [0, 1]$ be a Vitali set, and define $f: \mathbb{R} \rightarrow \mathbb{R}$ by

$$f(x) = \begin{cases} x + 1, & x \in V \\ -|x|, & x \notin V \end{cases}$$

By construction $\{f > 0\} = V$, which is nonmeasurable. Thus, f is nonmeasurable. On the other hand, for every $t \in \mathbb{R}$ the set $f^{-1}(t)$ is finite and therefore measurable. Indeed, if t is negative, $f(x) = t$ holds for at most two values of x ; and when $t \geq 0$, there is at most one such value. \square

[Remark: If we wanted to construct such an example on \mathbb{R}^n for every n , one way is to let

$$f(x_1, \dots, x_n) = \begin{cases} x_1 + 1, & \forall i \ x_i \in V; \\ -|x_1|, & \text{otherwise} \end{cases}$$

Then $\{f > 0\} = V^n$ which is nonmeasurable, because on one hand, $V^n + \mathbb{Q}^n = \mathbb{R}^n$ forces $|V^n|_e > 0$; on the other, $V^n + (\mathbb{Q} \cap [0, 1])^n$ is a bounded set containing infinitely many copies of V^n , which makes it impossible to have $|V^n| > 0$.

For every $t \in \mathbb{R}$, the preimage $f^{-1}(t)$ consists at most two hyperplanes of the form $\{x\} \times \mathbb{R}^{n-1}$. So it is covered by countably many sets of the form $\{x\} \times [-j, j]^{n-1}$, $j \in \mathbb{N}$. Here $|\{x\} \times [-j, j]^{n-1}| = 0$ because this set is contained in a box of dimensions $(\epsilon, 2j, \dots, 2j)$ whose volume can be arbitrarily small. In conclusion, $|f^{-1}(t)| = 0$ for every t . Thus f is measurable.]

Solution Problem 15. Suppose that $f: \mathbb{R} \rightarrow \mathbb{R}$ is measurable, and $g: \mathbb{R} \rightarrow \mathbb{R}$ is continuously differentiable with $g' > 0$ everywhere. Prove that $f \circ g$ is measurable.

Proof. By the Mean Value Theorem, g is strictly increasing; therefore, it has an inverse $h = g^{-1}$. By the Inverse Function Theorem, the inverse function h is also continuously differentiable.

Given $a \in \mathbb{R}$, consider the set $A = \{x : f(g(x)) > a\}$. It can be written as $\{x : g(x) \in B\}$, where $B = \{f > a\}$ is measurable; that is, $A = h(B)$. By Problem 9, the image of a measurable set under a continuously differentiable function is measurable. Therefore, A is measurable. \square

Solution Problem 16.

- (a) Suppose $f : \mathbb{R}^n \rightarrow \mathbb{R}$ is a continuous function such that f^2 is measurable. Prove that f is measurable.
- (b) Prove that the statement in (a) is false if f is not assumed continuous.

Proof.

- (a) Since f is continuous, it is measurable.
- (b) Let $n = 1$, let V be a Vitali set, and define $f(x) = 1$ when $x \in V$ and $f(x) = -1$ when $x \notin V$. Then $f^2 \equiv 1$ is measurable, being continuous. But $\{f > 0\} = V$ is not a measurable set, so f is not measurable. \square

Solution Problem 17. Suppose that $f : E \rightarrow \mathbb{R}$ is a measurable function, where $E \subset \mathbb{R}^n$ is measurable.

- (a) Prove that there exists a Borel set $H \subset E$ such that the restriction $f|_H$ is Borel measurable and $|E \setminus H| = 0$.
- (b) If, in addition, E is a Borel set, prove that there exists a Borel measurable function $g : E \rightarrow \mathbb{R}$ such that $f = g$ a.e.. [Hint: For part (a), take a countable union of closed sets obtained from Lusin's Theorem.]

Solution Problem 18. Suppose $\phi : [0, \infty) \rightarrow [0, \infty)$ is a function such that $\phi(t) \rightarrow 0$ as $t \rightarrow \infty$. Consider a sequence of measurable functions $\{f_k\}$, $f_k : \mathbb{R}^n \rightarrow \mathbb{R}$, such that $|f_k(x)| \leq \phi(|x|)$ for every k , and $f_k \rightarrow f$ a.e.. Prove that the conclusion of Egorov's Theorem holds in this situation; that is, for every $\epsilon > 0$, there exists a closed set $E(\epsilon) \subset \mathbb{R}^n$ such that $|\mathbb{R}^n \setminus E(\epsilon)| < \epsilon$ and $f_k \rightarrow f$ uniformly on $E(\epsilon)$. [Hint: Follow the proof of Egorov's theorem.]

Solution Problem 19. Suppose that $f: \mathbb{R}^n \rightarrow \mathbb{R}$ is a function such that $f(\mathbb{R}^n)$ is countable and $f^{-1}(t)$ is measurable for every $t \in \mathbb{R}$. Prove that f is measurable.

Solution Problem 20. Prove that without the assumption “ $f(\mathbb{R}^n)$ is countable” the statement in Problem 19 would not be true.

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