# Measure Theory

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In math, we are first taught to solve simple equations like  $x^2 - 2x + 4 = 0$  for a certain number x, but in real world applications, we must now solve for some function f satisfying an equation

$$\mathcal{L}(f) = 0 \tag{1}$$

where  $\mathcal{L}$  is some operator on functions. This is usually difficult, and many times a solution does not exist. However, we can find approximate solutions, say

$$\mathcal{L}(f) = 1/2$$

$$\mathcal{L}(f) = 1/4$$

$$\mathcal{L}(f) = 1/8$$

$$\dots = \dots$$

and approximate the solution as

$$f = \lim_{n \to \infty} f_n \tag{2}$$

Given that this limit exists, we can usually define f pointwise using a point-wise limit

$$f(x) = \lim_{n \to \infty} f_n(x) \text{ for all } x \tag{3}$$

but the function in total is very ugly and not Riemann integrable. The classic non-Riemann integrable function is the

$$f(x) = \chi_{\mathbb{R} \setminus \mathbb{Q}}(x) := \begin{cases} 1 & \text{if } x \in \mathbb{R} \setminus \mathbb{Q} \\ 0 & \text{if } x \in \mathbb{Q} \end{cases}$$
 (4)

Since  $\mathbb{Q}$  is countable, we can enumerate  $\mathbb{Q} = \{q_n\}_{n=1}^{\infty}$  and define the sequence of functions

$$f_n = 1 - \chi_{\{q_j\}_{j=1}^n}(x) \tag{5}$$

that start off with the constant function 1 and then "removes" points in  $\mathbb{Q}$ , setting their image to 0. It is clear that since we are removing points, every function in the sequence has an integral (from 0 to 1) of 1, and therefore the integral of f should also be 1.

$$\int_{0}^{1} f_{n} dx = 1 \implies \int_{0}^{1} f dx = \int_{0}^{1} \lim_{n \to \infty} f_{n} dx = \lim_{n \to \infty} \int_{0}^{1} f_{n} dx \tag{6}$$

What is crucial for mathematicians to work with is the capability to take the limit from inside the integral to outside the integral. The problem is that f is not a Riemann integral function.

## Definition 0.1 (Riemann Integrable Function)

Given a function  $f:[0,1] \to \mathbb{R}$ , let us consider some partition of [0,1] into intervals  $P = \{I_0, I_1, \ldots, I_N\}$ , then, for each  $I \in P$ , we can take the supremum  $M_I = \sup_{x \in I} f(x)$  and infimum  $m_I = \inf_{x \in I} f(x)$  and bound f by the upper and lower Riemann sums.

$$\sum_{I \in P} m_I |I| \le \int_0^1 f \, dx \le \sum_{I \in P} M_I |I| \tag{7}$$

where |I| is the length of interval I. If we take all possible partitions, the bound should still hold.

$$m = \sup_{P} \left\{ \sum_{I \in P} m_I |I| \right\} \le \int_0^1 f \, dx \le \inf_{P} \left\{ \sum_{I \in P} M_I |I| \right\} = M \tag{8}$$

and if the lower bound is equal to the upper bound m = M, then the integral is this number and f is considered Riemann integrable.

Now since  $\mathbb{Q}$  is dense in  $\mathbb{R}$ , for every interval I in every partition P will have  $m_I=0$  and  $M_I=1$  for the Riemann function, meaning that the lower bound will always be 0 and the upper bound will always be 1. So,  $\int_0^1 \chi_{\mathbb{R}\setminus\mathbb{Q}}(x)$  can take on any value in [0,1], which isn't helpful. The fact that we can't integrate this really simple function is a problem. For nice functions, we can partition it so that the base of each Riemann rectangle is a nice interval, while the base of the Riemann function is an "interval with holes." The problem really boils down to measuring what the "length" of this set is. So the problem with the Riemann integral isn't the integral itself, but the fact that we can't give a meaningful size to the set  $\mathbb{R}\setminus\mathbb{Q}$ . Now mathematicians in the 19th century thought that as long as we solve this problem, we should be good to go, but Banach and Tarski proved that there exists sets that cannot be measured with their famous paradox, which says that you can take any set P, disassemble it into a finite set of pieces, and rearrange it (under isometry and translations) so that it has a different size than the original P. So, we have to exclude some sets that are not measurable. The collection of sets that we can measure is called the  $\sigma$ -algebra.

#### 1 Sigma Algebras

In here, we will develop a deeper formalism of set theory and topology, now that we have the tools of analysis.

## **Set-Theroetic Limits**

Let's talk about sequences of sets  $(A_n)_n$ .

## Definition 1.1 (Monotone Sequence)

A sequence of sets  $(A_n)_n$  is called

- 1. (strictly) increasing if  $A_n \subseteq A_{n+1}$ .
- 2. nondecreasing if  $A_n \subseteq A_{n+1}$ .
- 3. (strictly) decreasing if  $A_n \supseteq A_{n+1}$ .
- 4. nonincreasing if  $A_n \supseteq A_{n+1}$ .

## Definition 1.2 (Limsup and Liminf of Sets)

Given a sequence of sets  $(A_n)_n$ , the **limsup** and **liminf** of them can be defined in the equivalent

1. The **liminf** is the set of points that are missing in only a finite number of sets, and the **limsup** is the set of points that are in an infinite number of sets.

$$\liminf_{n \to \infty} A_n := \bigcup_{n=1}^{\infty} \bigcap_{m=n}^{\infty} A_m \tag{9}$$

$$\liminf_{n \to \infty} A_n := \bigcup_{n=1}^{\infty} \bigcap_{m=n}^{\infty} A_m$$

$$\limsup_{n \to \infty} A_n := \bigcap_{n=1}^{\infty} \bigcup_{m=n}^{\infty} A_m$$
(10)

2. The **liminf** and **limsup** are the set of points x where the liminf and limsup of the indicator function function evaluated at x equals 1.

$$\liminf_{n \to \infty} A_n := \{ x \in X \mid \liminf_{n \to \infty} \mathbb{1}_{A_n}(x) = 1 \}$$
(11)

$$\limsup_{n \to \infty} A_n := \left\{ x \in X \mid \limsup_{n \to \infty} \mathbb{1}_{A_n}(x) = 1 \right\} \tag{12}$$

Both liminf and limsup always exist for any sequence of sets.

## Proof.

DeMorgan's law.

## Lemma 1.1 (Monotonicity)

For any sequence of sets

$$\liminf_{n \to \infty} A_n \subseteq \limsup_{n \to \infty} A_n \tag{13}$$

## Lemma 1.2 (Complements)

$$\liminf_{n \to \infty} A_n = \left( \limsup_{n \to \infty} A_n^c \right)^c \tag{14}$$

Proof.

Definition 1.3 (Limit of Sets)

## 1.2 Borel Hierarchy

Definition 1.4 ( $F_{\sigma}$  Sets)

A  $F_{\sigma}$ -set is a subset of a topological space that is a countable union of closed sets.

Definition 1.5 ( $G_{\delta}$  Sets)

A  $G_{\delta}$ -set is a subset of a topological space that is a countable intersection of open sets.

Lemma 1.3 ()

The complement of a  $F_{\sigma}$  set is a  $G_{\delta}$  set.

## 1.3 Sigma Algebra

Now, given any set X, we can construct its power set  $2^X$ . But we can't naively just give a measure to every  $A \in 2^X$ , since for certain spaces, this causes nasty contradictions shown through the Banach-Tarski Paradox.<sup>1</sup> A nice set of subsets of X to work with is the  $\sigma$ -algebra of X.

Definition 1.6 ( $\sigma$ -Algebra)

A  $\sigma$ -algebra on a set X is a collection of subsets of X, denoted  $\mathcal{A} \subset 2^X$  that contains  $\emptyset$ , X itself, is stable under a countable union, and is stable under complementation. This pair  $(X, \mathcal{A})$  is called a measurable space.

## Lemma 1.4 (Additional Property of $\sigma$ -Algebras)

A commonly known property of any  $\sigma$ -algebra  $\mathcal{A}$  is that it is stable under countable intersections, too.

$$A_1, A_2, \dots, \in \mathcal{A} \implies \bigcap_{k=1}^{\infty} A_k \in \mathcal{A}$$
 (15)

Proof.

We can utilize the fact that

$$\bigcap_{k=1}^{\infty} A_k = X \setminus \bigcup_{k=1}^{\infty} A_k^c \tag{16}$$

A  $\sigma$ -algebra is similar to the topology  $\tau$  of topological space. Both  $\mathcal{A}$  and  $\tau$  require  $\emptyset$  and X to be in it. The three differences are that (i)  $\tau$  does not allow compelmentation, (ii)  $\tau$  allows any (even uncountable) union

Given any two bounded subsets A and B of  $\mathbb{R}^n$  where  $n \geq 3$ , both of which have a nonempty interior, there are partitions of A and B into a finite number of disjoint subsets,  $A = A_1 \cup \ldots \cup A_k$ ,  $B = B_1 \cup \ldots \cup B_k$ , such that  $A_i$  and  $B_i$  are congruent for every  $i \in [k]$ .

of sets (condition is strengthened), and (iii)  $\tau$  allows only finite intersection of sets (condition is weakened). Now in order to construct  $\sigma$ -algebras, the following theorems are useful since they allow us to construct  $\sigma$ -algebras from other  $\sigma$ -algebras. It turns out that the intersection of  $\sigma$ -algebras is a  $\sigma$ -algebra, but not for unions.

## Theorem 1.1 (Intersection of Sigma Algebras is a Sigma Algebra)

Let  $\{A_k\}$  be a family of  $\sigma$ -algebras of X. Then,  $\cap A_k$  is also a  $\sigma$ -algebra of X.

#### Proof.

Clearly,  $\emptyset$ , X is in  $\cap A_k$ . To prove complementation,

$$A \in \bigcap \mathcal{A}_k \implies A \in \mathcal{A}_k \ \forall k \implies A^c \in A_k \ \forall k \implies A^c \in \bigcap \mathcal{A}_k$$
 (17)

To prove countable union, let  $\{A_j\}_{j\in J}$  be some countable family of subsets in  $\cap A_k$ . Then,

$$A_{j} \in \bigcap \mathcal{A}_{k} \ \forall j \in J \implies A_{j} \in \mathcal{A}_{k} \ \forall k \forall j \implies \bigcup A_{j} \in \mathcal{A}_{k} \ \forall k \implies \bigcup A_{j} \in \bigcap \mathcal{A}_{k}$$
 (18)

This allows us to easily prove the following theorem, which just establishes the existence of  $\sigma$ -algebras.

## Theorem 1.2 (Unique Smallest Sigma Algebra)

Let  $F \subset 2^X$ . Then there exists a unique smallest  $\sigma$ -algebra  $\sigma(F)$  containing F, called the  $\sigma$ -algebra **generated** by F.

## Proof.

Let us denote  $\mathcal{M}$  as the set of all possible  $\sigma$ -algebras  $\mathcal{B}$  of X.  $\mathcal{M}$  is nonempty since it contains  $2^X$ . Then, the intersection

$$\bigcap_{\mathcal{B}\in\mathcal{M}}\mathcal{B}\tag{19}$$

is the unique smallest  $\sigma$ -algebra.

With this guarantee, we can now define what it means for a set of subsets to generate a  $\sigma$ -algebra.

## Definition 1.7 ( $\sigma$ -Algebra Generated by a Set)

Given a collection of sets  $\mathscr{C}$ , the  $\sigma$ -algebra **generated** by  $\mathscr{C}$  is the unique smallest  $\sigma$ -algebra containing  $\mathscr{C}$ , denoted  $\sigma(\mathscr{C})$ .

This gives us a convenient way to construct  $\sigma$ -algebras. The general method is to identify a collection of "important" subsets that we would like to be included in the  $\sigma$ -algebra, and then just generate it.

## Definition 1.8 (Borel $\sigma$ -algebra)

The **Borel**  $\sigma$ -algebra of a topological space  $(X, \mathscr{T})$  is the  $\sigma$ -algebra generated by the topology  $\mathscr{T}$ , denoted  $\mathcal{B}(X) := \sigma(\mathscr{T})$ . An element of the Borel algebra is called a **Borel set**.

Note that the Borel algebra contains:

- 1. all open sets,
- 2. all closed sets due to closure under complements,

- 3. all  $G_\delta$  sets due to closure under countable unions,
- 4. all  $F_{\sigma}$  sets due to closure under countable intersection.

## Definition 1.9 (Measure Space)

A measure set is a tuple (X, A), where X is an arbitrary space and A a  $\sigma$ -algebra.

## 2 Measures

The introduction of the  $\sigma$ -algebra seemed quite arbitrary, but bear with me as it will make sense very soon. In general, we want to define a measure  $\mu: 2^X \to [0, +\infty]$  that satisfies two properties.

- 1. Null empty set.  $\mu(\emptyset) = 0$ .
- 2. Countable Additivity. For all countable collections  $\{A_k\}_{k=1}^{\infty}$  of pairwise disjoint subsets  $A_k \subset 2^X$ ,

$$\mu\bigg(\bigsqcup_{k=1}^{\infty} A_k\bigg) = \sum_{k=1}^{\infty} \mu(A_k) \tag{20}$$

The first condition is important because it allows us to take finite disjoint unions. That is, since  $\mu(A_1 \cup A_2) = \mu(A_1 \cup A_2 \cup \emptyset \cup \ldots)$ , we have

$$\sum_{k=1}^{\infty} = \mu(A_1) + \mu(A_2) \tag{21}$$

Disjointness is clearly important since if it wasn't, then  $\mu(A) = \mu(A \cup A) = 2\mu(A)$ , which is absurd.

It turns out that this second property is highly restrictive, and in fact some measures cannot be even defined. But this is self-contradictory, as it turns out that that we can create partitions of weird sets and rearrange them to get paradoxes (the most famous being the Banach-Tarski paradox). Therefore, we need to find a certain subset  $\mathcal{A} \subset 2^X$  that is consistent with this definition of measure.

- 1. We want to define a function  $\mu^*: 2^X \to [0, +\infty]$  that has a slightly less restrictive form of property 2.<sup>2</sup> We should always be able to construct such a function, which we will call the *outer measure*.
- 2. Then, we want to use this outer measure to define sets that should like in  $\mathcal{A}$ . We call these measurable sets. It will turn out that  $\mathcal{A}$  must be a  $\sigma$ -algebra.
- 3. Finally, we take the restriction of the outer measure to only measurable sets, and this defines our measure:  $\mu = \mu^*|_{\Delta}$ .

3

## Definition 2.1 (Measure)

Given a measurable space  $(X, \mathcal{A})$ , a **measure** is a function  $\mu : \mathcal{A} \longrightarrow [0, +\infty]^a$  satisfying

- 1. Null empty set  $\mu(\emptyset) = 0$ .
- 2. Countable additivity: For all countable collections  $\{A_k\}_{k=1}^{\infty}$  of pairwise disjoint subsets  $A_k \in \mathcal{A}$ ,

$$\mu\left(\bigsqcup_{k=1}^{\infty} A_k\right) = \sum_{k=1}^{\infty} \mu(A_k) \tag{22}$$

Remember that we are allowed to take countable unions inside our  $\sigma$ -algebra, so this makes sense.

This immediately implies that given  $A, B \in \mathcal{A}$ , then  $A \subset B \implies \mu(A) \leq \mu(B)$ . The triplet  $(X, \mathcal{A}, \mu)$  is called a **measure space**.

 $<sup>^</sup>a$ We usually introduce this by taking the codomain to be either  $[0,+\infty]$  or  $(-\infty,+\infty)$ , which is the signed measure.

<sup>&</sup>lt;sup>2</sup>How we implement such a function is a different question, though.

<sup>&</sup>lt;sup>3</sup>Old but good explanataion: Now let's try to construct a measure  $\lambda$  on the Borel  $\sigma$ -algebra  $\mathcal{B}(\mathbb{R})$  that assigns length, i.e.  $\lambda([a,b]) = b-a$ . We will do so by constructing outer measures  $\lambda^* : 2^{\mathbb{R}} \longrightarrow \mathbb{R}$  that acts on the power set of  $\mathbb{R}$  s.t.  $\lambda^*([a,b]) = b-a$ . But this turns out to have its own problems and contradictions, so once we construct such a  $\lambda^*$ , we will "throw away" all the sets that don't behave nicely under  $\lambda^*$  and just use its restriction on the Borel algebra. It turns out that the sets that do behave well under  $\lambda^*$  is bigger than the Borel algebra, call it  $\mathcal{M}_{\lambda^*}$ . So, we have  $\mathcal{B}(\mathbb{R}) \subset \mathcal{M}_{\lambda^*} \subset 2^{\mathbb{R}}$ . We will do this in full generality in the following way. We take any space X and construct an outer measure  $\mu^*$  on its power set  $2^X$ . Then, we construct the  $\sigma$ -algebra of well-behaved sets  $\mathcal{M}_{\mu^*} \subset 2^X$ , and define our measure  $\mu$  on  $\mathcal{M}_{\mu^*}$ . When defining our outer measure, the condition that the outer measure of a disjoint union of subsets is equal to the sum of the outer measure of the subsets is a bit too restricting, so we use a softer condition.

Let's go through each of these three steps in detail.

## 2.1 Outer Measure

## Definition 2.2 (Outer Measure)

Given a space X, an **outer measure** is a function  $\mu^*: 2^X \to [0, +\infty]$  satisfying either the two properties.

- 1. Null Empty Set.  $\mu^*(\emptyset) = 0$ .
- 2. Countable Subadditivity. For arbitrary subset  $A, B_1, B_2, \ldots$ ,

$$A \subset \bigcup_{k=1}^{\infty} B_k \implies \mu(A) \le \sum_{k=1}^{\infty} \mu(B_k)$$
 (23)

or equivalently, the three properties.

- 1. Null Empty Set.  $\mu^*(\emptyset) = 0$ .
- 2. Monotonicity. If  $A, B \subset X$ , then

$$A \subset B \implies \mu^*(A) \le \mu^*(B)$$
 (24)

3. Countable Subadditivity. For any countable collection of subsets  $\{A_k\}$  of X,

$$\mu^* \left( \bigcup_k A_k \right) \le \sum_k \mu^* (A_k) \tag{25}$$

#### Proof.

Prove that the two definitions are equal.

Okay, so what are some examples of outer measures that we can define on  $\mathbb{R}$ , or in general  $\mathbb{R}^n$ ? Well one approach would be to generalize the concepts length/area/volume. Such that for "simple" sets A where we know what the area is, the outer measure of A should coincide with the area of A. Let's first start by defining what a "simple" set is.

## Definition 2.3 (Elementary Set)

An elementary set  $E \subset \mathbb{R}^n$  is defined recursively as follows.

- 1. An **interval**  $I \subset \mathbb{R}$  is one of the sets (a,b), [a,b), (a,b], [a,b] for  $a,b \in \mathbb{R}$ .
- 2. For n > 1, an elementary set  $E \subset \mathbb{R}^n$  is  $E = I_1 \times \ldots \times I_n$  for intervals  $I_1, \ldots, I_k$ .

## Definition 2.4 (Size)

The size of an elementary set  $E = I_1 \times ... \times I_n \subset \mathbb{R}^n$  is defined recursively as

- 1. The **length** of an interval I is  $\ell(I) = b a$ .
- 2. The size of E is  $s(E) = \prod_{i=1}^{n} (b_i a_i)$ .

## Definition 2.5 (Lebesgue Outer Measure)

Given any set  $A \subset \mathbb{R}$ , the **Lebesgue outer measure** is defined

$$\lambda^*(A) = \inf \left\{ \sum_{k=1}^{\infty} \ell(I_k) \mid A \subset \bigcup_{k=1}^{\infty} I_k \right\}$$
 (26)

Intuitively, it is just the infimum of the sums of lengths of the intervals that cover A.

<sup>a</sup>I use the notation  $\mu^*$  to represent general outer measures, and  $\lambda^*$  to represent specifically the Lebesgue outer measure.

It's a hard definition, but a natural one, since we're taking all these intervals and trying to make them as snug as possible to define the outer measure of an arbitrary set. As always, let's begin with the simplest case in the real line. The following definition suffices.

## Lemma 2.1 (Lebesgue Outer Measure is an Outer Measure)

The Lebesgue outer measure  $\lambda^*$  on  $\mathbb{R}$  is indeed an outer measure.

## Proof.

We prove the three properties. The first two are trivial. For the third, we wish to show that  $\lambda^*(\cup A_n) \leq \sum \lambda^*(A_n)$ . For each n, find a specific cover  $\{I_{n_k}\}_{k=1}^{\infty}$  of  $A_n$  such that it "just covers" enough (this is possible since  $\lambda^*$  is an infimum) such that for any  $\epsilon > 0$ ,

$$\sum_{l_k} \ell(I_{n_k}) - \frac{\epsilon}{2^k} \le \lambda^*(A_k) \tag{27}$$

Then,

$$\bigcup A_n \subset \bigcup_{n,k=1} I_{n_k} \implies \sum_{n,k=1} \ell(I_{n_k}) \le \sum_{n=1}^{\infty} \lambda^*(A_n) + \epsilon$$
 (28)

and since  $\epsilon$  is arbitrary, we are done.

The first condition is trivial. As for 2, if I have  $A \subset B \subset \mathbb{R}$  and have a covering of B, then I also have a covering of A, and so the infimum corresponding to the covering of B must be greater than or equal to the infimum of that corresponding to the covering of A. For 3, we want to prove that the outer measure of the union of  $A_k$ 's is less than or equal to the sum of the outer measures of the  $A_k$ 's. We pick  $\epsilon > 0$  and have some covering  $\{(a_i^k, b_i^k)_{i=1}^\infty \in C_{A^k}$ . So we have

$$\lambda^*(A_k) \le \sum_{j=1}^{\infty} b_j^k - a_j^k \tag{29}$$

We want the inequality to go the other way around, but we can't do that. But note that  $\lambda^*(A_k)$  is the infimum of all coverings  $\{(a_j^k, b_j^k)\}_{j=1}^{\infty}$  of  $A_k$ , and so we can choose a covering that is as close to  $\lambda^*(A_k)$ , and then add a term of  $\epsilon$  to  $\lambda^*(A_k)$  to make it greater than this covering. This is an important step of the proof that is used often!

$$\frac{\epsilon}{2^k} + \lambda^*(A_k) \ge \sum_{j=1}^{\infty} b_j^k - a_j^k \tag{30}$$

Now.

$$A = \bigcup_{k=1}^{\infty} A_k \subset \bigcup_{k=1}^{\infty} \bigcup_{i=1}^{\infty} (a_j^k, b_j^k)$$
(31)

and we can see that  $\{(a_j^k, b_j^k)\}_{j,k=1}^{\infty} \in C_A$  is a countable covering of A (since the countable union of a countable union is countable), implying that

$$\lambda^*(A) \le \sum_{k=1}^{\infty} \sum_{j=1}^{\infty} (b_j^k - a_j^k) \le \sum_{k=1}^{\infty} \left(\lambda^*(A_k) + \frac{\epsilon}{2^k}\right) = \epsilon + \sum_{k=1}^{\infty} \lambda^*(A_k)$$
(32)

and so setting  $\epsilon$  arbitrarily small we have  $\lambda^*(A) \leq \sum_{k=1}^{\infty} \lambda^*(A_k)$ .

Not only is it an outer measure; it also satisfies the property that we wanted, along with the bonus property of translation invariance!

## Theorem 2.1 (Lebesgue Outer Measure Coincides with Interval Length)

 $\lambda^*$  satisfies the property that for any interval  $I \subset \mathbb{R}$ ,  $\lambda^*(I) = S(I)$ .

## Proof.

Let I = [a, b]. Take  $I_1 = [a - \epsilon, b + \epsilon]$ .

- 1. For an upper bound, we see  $\lambda^*(I) \leq b a + 2\epsilon$ , where  $2\epsilon \to 0$ .
- 2. For a lower bound, suppose

$$\bigcup_{n=1}^{\infty} I_n \supset [a, b] \tag{33}$$

By Heine-Borel, we can extract a finite subcollection  $I_1, \ldots, I_n$  that still covers [a, b]. Under  $(a_k, b_k)$  so that  $a_k$  is increasing.

- (a) Consider  $(a_1, b_1)$ . If  $b_1 > b$ , we are done.
- (b) Otherwise,  $b_1 \in (a_2, b_2)$ . If  $b_2 > b$ , then

$$b_2 - a_2 + b_1 - a_1 \ge b_2 - a_1 > b - a \tag{34}$$

(c) If not, then we keep going until we get to  $(a_n, b_n)$ . If  $b_n > b$ , then

$$b_n - a_n + b_{n-1} - a_{n-1} + \ldots + b_1 - a_1 \ge b_n - a_1 > b - a \tag{35}$$

## Corollary 2.1 (Translation Invariance)

 $\lambda^*$  is translation invariant. That is, for any  $A \subset \mathbb{R}$ ,

$$\lambda^*(A) = \lambda^*(A+x) \tag{36}$$

where  $A + x := \{a + x \in \mathbb{R} \mid a \in A\}.$ 

#### Theorem 2.2 (Countable Sets have Outer Measure 0)

Any countable set of  $\mathbb{R}$  has Lebesgue outer measure 0.

## Proof.

Just enumerate  $A = \{x_1, \ldots\}$ . Then, we set  $I_k = (x_k - \frac{\epsilon}{2^k}, x_k + \frac{\epsilon}{2^k})$ . Then,

$$\sum_{k=1}^{\infty} \ell(I_k) = \epsilon \tag{37}$$

We can also generalize this further by introducing a increasing, continuous function  $F: \mathbb{R} \to \mathbb{R}$  and defining the outer measure to be

$$\lambda^*(A) = \inf_{C_A} \sum_{j=1}^{\infty} \left( F(b_j) - F(a_j) \right)$$
(38)

In  $\mathbb{R}^n$ , this construction is exactly the same, since we can take rectangular prisms, which we know the area/volume of, make a countable covering of some arbitrary set  $A \subset \mathbb{R}^n$ , and then find the infimum of the volume of this set. But we can't apply the outer measure on power sets since there exists some sets that do

not behave like how we want it to behave under a measure. For example, there exists disjoint  $A, B \subset (0,1)$  s.t.  $A \cup B = (0,1)$ , but  $\lambda^*(A) + \lambda^*(B) > 1$ .

## 2.2 Measurable Sets

## Definition 2.6 (Carathéodory's criterion)

Given outer measure  $\mu^*$  on X, a set  $E \subset X$  is called  $\mu^*$ -measurable if for every set  $A \subset X$ ,

$$\mu^*(A \cap E) + \mu^*(A \cap E^c) = m^*(A) \tag{39}$$

In general it says that no matter how nasty a subset A is, E should be nice enough that we can cut E into two pieces C and D. Due to the definition of the outer measure, we are guaranteed to have  $\mu^*(C \cup D) \le \mu^*(C) + \mu^*(D)$ . The sets with which this inequality is strict is not measurable, and the measurable sets specifically satisfy

- 1. equality
- 2. for countable sets.

One should note that in particular, if E is  $\mu^*$ -measurable and A is any set disjoint from E, then we must have

$$\mu^*(A \cup E) = \mu^*((A \cup E) \cap E) + \mu^*((A \cup E) \cap E^c)$$
(40)

$$= \mu^*(E) + \mu^*(A) \tag{41}$$

which solves a bit of the theorem on measures. In practice, we will often prove that  $\mu^*(A \cap E) + \mu^*(A \cap E^c) \le m^*(A)$ , since the properties of outer measure implies  $\ge$ .

#### Example 2.1 ()

Take  $X = \mathbb{R}$  and have  $B = (-\infty, b]$ . Then  $B^c = (b, \infty)$ , and B divides  $\mathbb{R}$  into a right side and a left side. If we take any subset  $A \subset \mathbb{R}$ , then B is nice enough to divide A into a left and a right side.

Now we want to establish some nice properties.

#### Theorem 2.3 (Outer Measure 0 Sets are Measurable)

For any outer measure  $\mu^*$  on X,  $E \subset X$  with  $\mu^*(E) = 0$  implies that E is  $\mu^*$ -measurable.

## Proof.

Take any A. Then  $(A \cap E) \subset E$  and  $(A \cap E^c) \subset A$ . So by monotonicity,

$$\mu^*(A \cap E) + \mu^*(A \cap E^c) < \mu^*(E) + \mu^*(A) = \mu^*(A) \tag{42}$$

and this by definition means that E is measurable.

Now let's talk about constructing measurable sets.

## Theorem 2.4 (Finite Unions are Outer Measurable)

A finite union of  $\mu^*$ -measurable sets is  $\mu^*$ -measurable.

#### Proof.

It suffices to prove for  $E_1, E_2$ , and the rest follows by induction. Fix any A. Then

$$\mu^*(A) = \mu^*(A \cap E_1) + \mu^*(A \cap E_1^c) \tag{43}$$

$$= \mu^*(A \cap E_1) + \mu^*((A \cap E_1^c) \cap E_2) + \mu^*((A \cap E_1^c) \cap E_2^c)$$
(44)

But

$$(A \cap E_1^c) \cap E_2^c = A \cap (E_1 \cup E_2)^c \tag{45}$$

$$(A \cap E_1^c) \cap E_2 = (A \setminus E_1) \setminus E_2 \tag{46}$$

So, 
$$(A \cap E_1) \cup ((A \setminus E_1) \cap E_2) = A \cap (A \cap (E_1 \cup E_2)^c)$$
.

So we have proved that the set of all measurable sets is closed under finite unions. By definition it works for finite intersections. This makes it into an *algebra*, but we want to upgrade this to a  $\sigma$ -algebra by proving closure under *countable* unions. We will need the lemma.

## Lemma 2.2 ()

Suppose  $E_1, \ldots, E_n$  are disjoint. Then,

$$\mu^* \left( \bigcup_{j=1}^n E_j \right) = \sum_{j=1}^n \mu^*(E_j) \tag{47}$$

## Proof.

We already did this for 2 sets, and just use induction.

Now we prove lemma, which is more general (arbitrary intersections than finite?).

#### Lemma 2.3 ()

Suppose A is any set,  $E_i$  disjoint and measurable. Then,

$$\mu^* \left( A \cap \left( \bigcup_{j=1}^n E_j \right) \right) = \sum_{j=1}^n \mu^* (A \cap E_j)$$

$$\tag{48}$$

## Proof.

By induction, n = 1 is true. Then,

$$\mu^* \left( A \cap \left( \bigcup_{j=1}^n E_j \right) \right) = \mu^* \left( \left( A \cap \left( \bigcup_{j=1}^n E_j \right) \right) \cap E_n \right) + \mu^* \left( \left( A \cap \left( \bigcup_{j=1}^n E_j \right) \right) \cap E_n^c \right)$$
(49)

$$= \mu^* (A \cap E_n) + \mu^* \left( A \cap \left( \bigcup_{i=1}^{n-1} E_i \right) \right)$$

$$\tag{50}$$

$$=\sum_{j=1}^{n}\mu^*(A\cap E_j)\tag{51}$$

by the induction hypothesis.

## Theorem 2.5 (Countable Unions are Outer Measurable)

Suppose  $E_1, E_2, \ldots$  are a countable collection of measurable sets. Then,  $E = \bigcup_{j=1}^{\infty} E_j$  is measurable.

#### Proof.

They key is to look at disjoint sets. WLOG, one can assume  $E_j$  are disjoint. Indeed, we can define new sets

$$E_n' := E_n \setminus \left(\bigcup_{j=1}^{n-1} E_j\right) \tag{52}$$

that are measurable, with  $\bigcup E'_n = \bigcup E_n$ . Now, fix any set A. Define sets  $F_n = \bigcup_{j=1}^n E_j$ . Then,  $\mu^*(A) = \mu^*(A \cap F_n) + \mu^*(A \cap F_n^c)$ . Then,  $F_n^c \supset E^c \implies \mu^*(A \cap F_n^c) \ge \mu^*(A \cap E^c)$ . Through the previous lemma, we have

$$\mu^*(A \cap F_n) = \mu^* \left( \bigcup_{j=1}^n (A \cap E_j) \right) = \sum_{j=1}^n \mu^*(A \cap E_j)$$
 (53)

Then,

$$\mu^*(A) \ge \sum_{j=1}^n \mu^*(A \cap E_j) + \mu^*(A \cap E^c)$$
(54)

for every n, therefore also with  $\infty$ . But

$$\sum_{j=1}^{\infty} \mu^*(A \cap E_j) \ge \mu^*(A \cap E) \tag{55}$$

If follows that  $\mu^*(A) \ge \mu^*(A \cap E) + \mu^*(A \cap E^c)$ .

## Corollary 2.2 (Measurable Sets form a $\sigma$ -Algebra)

The set of all  $\mu^*$ -measurable sets of X form a  $\sigma$ -algebra.

With this, we can construct a lot of measurable sets.

## Lemma 2.4 (Sets of Measure 0 have no Effect)

Suppose  $\mu^*(E) = 0$  and A is any set. Then,  $\mu^*(A \cup E) = \mu^*(A)$ .

## Proof.

We have

$$\mu^*(A \cup E) = \underbrace{\mu^*((A \cup E) \cap E)}_{=0} + \mu^*\underbrace{((A \cup E) \cap E^c)}_{\subseteq A} \le \mu^*(A) \le \mu^*(A)$$
 (56)

But  $A \cup E \supset A$ , so  $\mu^*(A \cup E) = \mu^*(A)$ .

So we can always drop an outer-measure 0 set and it won't affect the outer measure of the original set.

## Theorem 2.6 ()

Every interval  $(a, +\infty)$  is measurable.

#### Proof.

Take any set A, and WLOG  $a \notin A$  (since we can take the point out without affecting outer measure). Suppose  $\{I_k\}_{k=1}^{\infty}$  is a cover of A s.t.

$$\mu^* > \left(\sum_{k=1}^{\infty} \ell(I_k)\right) - \epsilon \tag{57}$$

Then,

1.  $I_k' := I_k \cap (a, +\infty)$  will cover  $A_1 = A \cap (a, +\infty)$ , and 2.  $I_k'' := I_k \cap (-\infty, a)$  will cover  $A_2 = A \cap (-\infty, a)$ . Therefore,  $\mu^*(A_1) \leq \sum_k \ell(I_k')$ ,  $\mu^*(A_2) \leq \sum_k \ell(I_k'')$ . Also,

$$\ell(I_k) = \ell(I_k') + \ell(I_k'') \implies \mu^*(A_1) + \mu^*(A_2) \le \sum_k \ell(I_k) \le \mu^*(A) + \epsilon$$
 (58)

for every  $\epsilon > 0$ . Since this is true for every  $\epsilon > 0$ , we are done.

## Theorem 2.7 ( $\lambda^*$ -measurable Sets)

TFAE in  $\mathbb{R}$  with the Lebesgue outer measure. E is measurable.

- 1.  $\forall \epsilon > 0, \exists \text{ open set } O \supset E \text{ s.t. } \mu(O \setminus E) \leq \epsilon.$
- 2.  $\forall \epsilon > 0, \exists \text{ closed set } F \subset E \text{ s.t. } \mu^*(E \setminus F) < \epsilon.$
- 3.  $\exists$  a  $G_{\delta}$  set G s.t.  $E \subset G$  and  $\mu^*(G \setminus E) = 0$ .
- 4.  $\exists$  a  $F_{\sigma}$  set F s.t.  $F \subset E$  and  $\mu^*(E \setminus F) = 0$ .

So essentially, we can construct measurable sets with "nice" sets.

## Proof.

Listed.

1.

For  $\mathbb{R}$ , we can create our Lebesgue outer measure  $\lambda^*$  on it, which generates the Lebesgue  $\sigma$ -algebra  $\mathcal{M}_{\lambda^*}$ . This turns out to be bigger than the Borel  $\sigma$ -algebra  $\mathcal{B}(\mathbb{R})$ , but there is little difference in which one we choose when we actually integrate.

## Theorem 2.8 ()

A set  $E \subset \mathbb{R}$  is Lebesgue measurable implies that it is also Borel measurable.

$$\mathcal{B}(\mathbb{R}) \subset \mathcal{M}_{\lambda^*} \subset 2^{\mathbb{R}} \tag{59}$$

## Lemma 2.5 ()

If  $E \subset \mathbb{R}$  and  $\lambda^*(E) = 0$ , then  $E \in \mathcal{M}_{\lambda^*}$ , i.e. E is Lebesgue outer-measurable.

#### Proof.

We must prove that E satisfies the Carathéodory's criterion. For all  $E \subset \mathbb{R}$ , we know that  $\lambda^*(A) \leq$  $\lambda^*(A \cap E) + \lambda^*(A \cap E^c)$  by definition of outer measure. Now, since  $\lambda^*(E) = 0$  and  $A \cap E \subset E$ , this means that  $\lambda^*(A \cap E) = 0$  also. Furthermore,  $A \cap E^c \subset A$ , meaning that  $\lambda^*(A) \geq \lambda^*(A \cap E^c)$ , and we get

$$\lambda^*(A) \ge \lambda^*(A \cap E) + \lambda^*(A \cap E^c) \tag{60}$$

which proves equality.

## 2.3 Measures

## Theorem 2.9 ()

The restriction of an outer measure  $\mu^*$  to the set of all  $\mu^*$ -measurable sets  $\mathcal{A}$ , denoted  $\mu = \mu^*|_{\mathcal{A}}$ , is measurable.

Now there are nice properties that we want Lebesgue measures to have: completeness, regularity, and translation invariance.

- 1. Completeness: Given sets  $A \subset B \subset C$  with  $\mu(A) = \mu(C)$  and  $A, C \in A$ , this implies that  $B \in A$ . This basically says that if you a set that is squeezed in between two measurable sets of equal measure, then the middle set will also be measurable.
- 2. Regularity: Given sets  $A \subset B \subset C$ , regularity talks about whether I can approximate B well. Most nice measures have this property.

$$\sup_{A \text{ compact}} \mu(A) = \mu(B) = \inf_{C \text{ open}} \mu(C)$$
 (61)

3. Translation invariance: Lebesgue measure is translation invariant.  $\mu(x+A) = \mu(A)$  for all  $x \in \mathbb{R}^n$  on  $\mathcal{B}(\mathbb{R}^n)$ .

## Definition 2.7 (Almost Everywhere)

Given a measure space  $(X, \mathcal{A}, \mu)$ , a subset  $A \in \mathcal{A}$  is said to be a  $\mu$ -null set if  $\mu(A) = 0$ . If some property holds for all points  $x \in X$  except on a null set, then we say that the property holds **almost** everywhere.

## Example 2.2 (Rational Function)

The function  $f(x) = \frac{1}{\sqrt{|x|}}$  is less than  $\infty$  almost everywhere.

Let us first look into some properties of measures, which all seem natural.

#### Theorem 2.10 ()

If 
$$A_1 \subset A_2 \subset A_3 \subset \ldots$$
, then

$$\mu\bigg(\bigcup_{k=1}^{\infty} A_k\bigg) = \lim_{k \to \infty} \mu(A_k) \tag{62}$$

## Proof.

This is the first time we introduce limits. With the fact that  $\mu(A_k)$  must be nondecreasing, we can use real analysis and see that it is bounded by  $\infty$ , meaning that it must have a limit. But why does

this limit equal to the left hand side? We can see that

$$\mu\left(\bigcup_{k=1}^{\infty} A_k\right) = \mu(A_1) + \sum_{k=2}^{\infty} \mu(B_k) \tag{63}$$

$$=\mu(A_1) + \lim_{k \to \infty} \sum_{k=2}^{\infty} \mu(B_k) \tag{64}$$

$$= \lim_{k \to \infty} \mu(A_1 \cup B_2 \cup \dots B_k) = \lim_{k \to \infty} \mu(A_k)$$
 (65)

where  $B_k = A_k \setminus A_{k-1}$ .

Now a similar theorem, but with a little twist to it.

## Theorem 2.11 ()

If  $A_1 \supset A_2 \supset A_3 \supset \dots$ , then

$$\mu\bigg(\bigcap_{k=1}^{\infty} A_k\bigg) = \lim_{k \to \infty} \mu(A_k) \tag{66}$$

if  $\mu(A_1) < \infty$ .

## Proof.

The  $\mu(A_1) < \infty$  is a necessary condition, since if we take  $A_k = [k, \infty)$  on the real number line, then we have  $\bigcap_{k=1}^{\infty} A_k = \emptyset$ , but the limit of the measure is  $\infty$ . Well we can define  $B_k = A_k \setminus A_{k+1}$  and write  $\bigcap_{k=1}^{\infty} A_k = A_1 \setminus \bigcup_{k=1}^{\infty} B_k$ , which means that

$$\mu\left(\bigcap_{k=1}^{\infty} A_k\right) = \mu\left(A_1 \setminus \bigcup_{k=1}^{\infty} B_k\right)$$

$$= \mu(A_1) - \mu\left(\bigcup_{k=1}^{\infty} B_k\right)$$

$$= \mu(A_1) - \sum_{k=1}^{\infty} \mu(B_k)$$

$$= \mu(A_1) - \lim_{K \to \infty} \sum_{k=1}^{K} \mu(B_k)$$

$$= \lim_{K \to \infty} \left(\mu(A_1) - \sum_{k=1}^{K} \mu(B_k)\right)$$

$$= \lim_{K \to \infty} \mu\left(A_1 \setminus \bigcup_{k=1}^{K} B_k\right) = \lim_{K \to \infty} \mu(A_K)$$

Now the first line uses the fact that if  $A \subset B$ , then  $\mu(B \setminus A) + \mu(A) = \mu(B)$ , and with the further assumption that  $\mu(A) < \infty$ , we can subtract on both sides like we do with regular arithmetic.

## 3 Measurable Functions

So far, we have defined measurable sets, constructed the Lebesgue measure, and shown that Lebesgue measurable sets can be approximated by nice open sets. Now, let's talk about measurable functions.

## Definition 3.1 (Measurable Function)

Given a measurable space  $(X, \mathcal{A}), f: (X, \mathcal{A}) \longrightarrow \mathbb{R}$  is **measurable** if

$$f^{-1}(A) \in \mathcal{A} \text{ for all } A \text{ open}$$
 (67)

Note that if we take  $\mathbb{R}^n$ , it can have either its Borel  $\sigma$ -algebra  $\mathcal{B}(\mathbb{R}^n)$  or its Lebesgue  $\sigma$ -algebra  $\mathcal{M}_{\lambda^*}$ . Therefore, a function  $f: \mathbb{R}^n \longrightarrow \mathbb{R}$  is said to be **Lebesgue measurable** (**Borel measurable**) if for every  $E \in \mathcal{B}(\mathbb{R})$ ,  $f^{-1}(E) \in \mathcal{M}_{\lambda^*}$  ( $f^{-1}(E) \in \mathcal{B}(\mathbb{R}^n)$ ). Since  $\mathcal{B}(\mathbb{R}^n) \subset \mathcal{M}_{\lambda^*}$ , all Borel measurable functions are Lebesgue measurable. It follows that any continuous function  $f: \mathbb{R}^n \longrightarrow \mathbb{R}$  is Borel (and hence Lebesgue measurable).

#### Theorem 3.1 ()

Let  $f: E \subset \mathbb{R} \to \mathbb{R} \cup \{\pm \infty\}$  and E be measurable. Then, TFAE

- 1.  $\forall c \in \mathbb{R}, \{x \in E \mid f(x) > c\}$  is measurable.
- 2.  $\forall c \in \mathbb{R}, \{x \in E \mid f(x) \geq c\}$  is measurable.
- 3.  $\forall c \in \mathbb{R}, \{x \in E \mid f(x) < c\}$  is measurable.
- 4.  $\forall c \in \mathbb{R}, \{x \in E \mid f(x) \leq c\}$  is measurable.
- 5. f is Lebesgue measurable.

Furthermore, if any of these hold, then also  $\{x \in E \mid f(x) = c\}$  is measurable for all c (but not the converse!).

## Proof.

We know that  $(1) \iff (4)$  and  $(2) \iff (3)$  by taking complements. We prove  $(1) \iff (2)$ .

 $1. (1) \Longrightarrow (2).$ 

$$\{x \in E \mid f(x) \ge c\} = \bigcap_{k=1}^{\infty} \{x \in E \mid f(x) > c - \frac{1}{k}\}$$
 (68)

 $2. (2) \Longrightarrow (1).$ 

$$\{x \in E \mid f(x) > c\} = \bigcup_{k=1}^{\infty} \{x \in E \mid f(x) \ge c + \frac{1}{k}\}$$
 (69)

For (5), we know that

- 1. (5)  $\implies$  (1) is trivial, since open intervals are open sets.
- 2.  $(1) \implies (5)$ . Any open set is a countable union of disjoint open intervals, and so let

$$U = \bigcup_{k=1}^{\infty} I_k, \qquad I_k = (a_k, b_k) = \underbrace{(-\infty, b_k)}_{B_k} \cap \underbrace{(a_k, +\infty)}_{A_k}$$
 (70)

Therefore,

$$f^{-1}(U) = \bigcup_{k=1}^{\infty} f^{-1}(B_k \cap A_k) = \bigcup_{k=1}^{\infty} \{f^{-1}(B_k) \cap f^{-1}(A_k)\}$$
 (71)

which is measurable since countable union/intersections are measurable (by definition of  $\sigma$ -algebra).

For the final implication, we can use (2) and (4) to get

$$\{x \in E \mid f(x) = c\} = \{x \in E \mid f(x) \le c\} \cup \{x \in E \mid f(x) \ge c\} \tag{72}$$

This implies that monotonic functions on  $\mathbb{R}$  are measurable. For example, take  $f:[a,b] \longrightarrow \mathbb{R}$  that is nondecreasing. Then, we would like to show that the preimage of every half-interval  $(-\infty,t)$  under f is in  $\mathcal{B}(\mathbb{R})$ . Well if we assume  $f(a) \geq t$ , then  $f(x) > t \ \forall t \in [a,b]$ , and so its preimage is  $\emptyset$ . If f(a) < t, having f(b) < t also leads to the preimage being [a,b] (which is the entire space and is in  $\mathcal{B}(\mathbb{R})$ ), and having f(b) > t implies that the preimage is  $[a,f^{-1}(t)]$ .

## Theorem 3.2 (Monotonic Functions are Measurable)

If  $f: E \subset \mathbb{R}$  is monotone, then f is measurable.

#### Proof.

Maybe on midterm lol.

## Theorem 3.3 (Continuous Functions are Measurable)

If  $f: X \to \mathbb{R}$  is continuous, then it is measurable.

#### Proof.

If f is continuous, then  $f^{-1}(O) = U \cap X$  for every open O with open U.

## Theorem 3.4 (Measurable Partition Induces Measurable Restrictions of Functions)

Suppose  $f: E \subset \mathbb{R} \to \mathbb{R} \cup \{\pm \infty\}$  with E measurable, and let g be some other function.

- 1. If f is measurable on E and g(x) = f(x) a.e. for  $x \in E$ , then g is measurable on E.
- 2. If  $D \subset E$  is measurable, then f measurable on E iff f measurable on D and on  $E \setminus D$ .

## Proof.

Listed.

1. Consider  $S = \{x \in E \mid g(x) < c\}$ . Let  $A \subset E$  be the set where g(x) = f(x), with  $m(E \setminus A) = 0$ . Then,

$$S = (\{x \in E \mid g(x) < c\} \cap (E \setminus A)) \cup (\{x \in E \mid f(x) < c\} \cap A)$$

$$(73)$$

where the first term is measure 0 by monotonicity with  $E \setminus A$ ,  $m(E \setminus A) = 0$ , and the second term is measurable since  $A = E \setminus (E \setminus A)$ . So, S is measurable.

2. For the second, we have

$$\{x \in E \mid f(x) < c\} = \{x \in D \mid f(x) < c\} \cup \{x \in E \setminus D \mid f(x) < c\}$$
(74)

$$(75)$$

$$\implies \{x \in D \mid f(x) < c\} = \{x \in E \mid f(x) < c\} \cap D \tag{76}$$

So there is some notion of robustness. You preserve measurability if you split the domain in a "measurable way."

The following theorem is useful, since we don't want to manually check measurability of every single new function we create.

## Theorem 3.5 (Arithmetic on Measurable Functions)

Given measurable functions  $f,g:E\subset\mathbb{R}\to\mathbb{R}$ , the following standard operations on them create new measurable functions:

- 1.  $\alpha f$  is measurable for all  $\alpha \in \mathbb{R}$ .
- 2. f + g is measurable
- 3.  $f \cdot g$  is measurable
- 4. f/g is measurable on  $\{x \mid g(x) \neq 0\}$

#### Proof.

WLOG, we can assume f, g are finite everywhere since changing these values to finite values over a set of measure 0 doesn't affect measurability.

1. If  $\alpha = 0$ , this is trivially true. If not, then

$$\{x \in E \mid (\alpha f)(x) < c\} = \{x \in E \mid f(x) < \frac{c}{\alpha}\}$$
 (77)

2. Suppose  $f(x) + g(x) < c \iff f(x) < c - g(x) \iff \exists q \in \mathbb{Q} \text{ s.t. } f(x) < g < c - g(x).^a \text{ Then,}$ 

$$\{x \in E \mid f(x) + g(x) < c\} = \bigcup_{q \in \mathbb{Q}} (\{x \in E \mid f(x) < q\} \cap \{x \in E \mid g(x) < c - q\})$$
 (78)

which is a countable union of measurable sets, and is measurable.

3. We use a nice trick from analysis.

$$fg = \frac{1}{4} ((f+g)^2 - (f-g)^2)$$
 (79)

and so it suffices to prove that h measurable implies  $h^2$  measurable. For  $c \geq 0^b$ , we have

$$\{x \in E \mid h^2(x) > c\} = \{x \in E \mid h(x) > \sqrt{c}\} \cup \{x \in E \mid h(x) < -\sqrt{c}\}$$
(80)

4.

#### Theorem 3.6 (Finite Min/Max of Measurable Functions are Measurable)

If  $f_1, \ldots, f_n : E \subset \mathbb{R} \to \mathbb{R}$  are measurable, then so are  $\max_k f_k$  and  $\min_k f_k$ .

#### Proof.

We can prove by induction, but this is still a one-liner. For maximum,

$$\{x \in E \mid (\max_{k} f_{k})(x) > c\} = \bigcup_{k=1}^{n} \{x \in E \mid f_{k}(x) > c\}$$
(81)

and for the minimum,

$$\{x \in E \mid (\min_{k} f_{k})(x) > c\} = \bigcap_{k=1}^{n} \{x \in E \mid f_{k}(x) > c\}$$
(82)

<sup>&</sup>lt;sup>a</sup>The reason we want to introduce rationals is that we want to take advantage of countability.

<sup>&</sup>lt;sup>b</sup>We only need to consider this case since  $h^2$  is always nonnegative and so c < 0 would mean preimage is empty set.

## 3.1 Composition of Measurable Functions

## Example 3.1 (Composition of Two Functions need not be Measurable)

Recall last time, we built a continuous strictly increasing function  $\psi(x)$  and showed that there exists a measurable A s.t.  $\psi(A)$  is not measurable. Let's extend  $\psi$  to all  $\mathbb{R}$  and keep it strictly increasing. Let  $\chi_A$  be the characteristic function of A. Consider  $f = \chi_A \circ \psi^{-1}$ , and take the preimage of  $(1/2, +\infty)$  under  $\psi$ .

$$f^{-1}((1/2, +\infty)) = \{x \mid \psi^{-1}(x) \in A\} = \{x \in \psi(A)\} = \psi(A)$$
(83)

which we have proven that there exists some A s.t.  $\psi(A)$  is not measurable.

So this is bad news, but we have a compromise.

## Theorem 3.7 (Composition of Measurable then Continuous is Measurable)

Suppose g is measurable on E, f is continuous on  $\mathbb{R}$ . Then,  $f \circ g$  is measurable.

#### Proof.

Take any open O. Then,

$$(f \circ g)^{-1}(O) \iff g(x) \in f^{-1}(O) \tag{84}$$

where  $f^{-1}(O)$  is open, which implies measurable.

So we get much more results, like that |f| or  $|f|^p$  is measurable if f is measurable.

## 3.2 Sequences of Measurable Functions

Let's compare continuous functions and measurable functions. In terms of composition, continuity is a little more robust since we can compose continuous functions to get continuous functions. Meanwhile, we know that measurable functions don't necessarily compose to measurable functions. The relation is reversed when we talk about convergence. Remember from analysis that

- 1.  $f_n \to f$  pointwise over E if  $f_n(x) \to f(x)$  for all  $x \in E$ , and we call it almost sure convergence if for all  $x \in A \subset E$  with  $m(E \setminus A) = 0$ .
- 2.  $f_n \to f$  uniformly over E if  $\forall \epsilon > 0$ ,  $\exists N \in \mathbb{N}$  s.t.  $\forall n > N \ \forall x \in E$ ,  $|f_n(x) f(x)| < \epsilon$ .

If you have uniform convergence, this is great since the uniform limit of continuous (Riemann integrable) functions is continuous (Riemann integrable). However, the pointwise limit of continuous (Riemann integrable) functions may fail to be continuous (Riemann integrable). It turns out that measurability is preserved through almost sure convergence.

#### Theorem 3.8 (Almost Sure Convergence of Measurable Functions are Measurable)

Suppose  $f_n$  are measurable on E and  $f_n \to f$  a.e. on E. Then, f is measurable.

So though continuous functions are more robust w.r.t. composition, measurable functions are more robust w.r.t. convergence.

## 3.3 Approximation of Measurable Functions

So if we throw out a small set, we can approximate measurable functions with continuous functions. Consider a generalization of step functions called *simple functions*.

## Definition 3.2 (Simple Functions)

For  $A \subset X$  (any subset, not just in some  $\sigma$ -algebra), the **characteristic**, or **indicator function** of A is the function  $\chi_A : X \longrightarrow \mathbb{R}$  defined

$$\chi_A(x) = \begin{cases} 1 & \text{if } x \in A \\ 0 & \text{if else} \end{cases}$$
 (85)

A function  $\phi : \mathbb{R} \longrightarrow \mathbb{R}$  is called a **simple function** if it is a finite linear combination of characteristic functions.

$$\phi = \sum_{i=1}^{n} a_i \chi_{A_i} \tag{86}$$

## Lemma 3.1 (Luzin, for Simple Functions)

Suppose  $f: E \subset \mathbb{R} \to \mathbb{R}$  is simple. Then  $\forall \epsilon > 0$ ,  $\exists$  a closed  $F \subset E$ ,  $g \in C(\mathbb{R})$ , s.t.  $f|_F = g|_F$  and  $m(E \setminus F) < \epsilon$ .

## Proof.

## Theorem 3.9 (Luzin)

Suppose  $f: E \subset \mathbb{R} \to \mathbb{R}$  is measurable. Then  $\forall \epsilon > 0, \exists$  a closed  $F \subset E, g \in C(\mathbb{R}), \text{ s.t. } f|_F = g|_F$  and  $m(E \setminus F) < \epsilon$ .

## Proof.

Assume  $m(E) < \infty$ . If it is  $\infty$ , then we can divide the set into countable sets, each with finite measure, and we can do it for  $\epsilon/2^n$ . So it suffices for finite measure. Suppose  $f_n$  are simple and  $f_n \to f$  pointwise on E. By the lemma, we can find closed  $F_n \subset E$  s.t.  $m(E \setminus F_n) < \epsilon/2^{n+1}$  and  $g_n \in C(\mathbb{R})$ , with  $g_n|_{F_n} = f_n|_{F_n}$ . Also, by Egoroff, we can find  $F_0$  s.t.  $f_n$  converges uniformly on  $F_0$ , and  $m(E \setminus F_0) < \epsilon/2$ . Define

$$F = \bigcap_{n=0}^{\infty} F_n, \qquad E \setminus F = \bigcup_{n=0}^{\infty} E \setminus F_n \tag{87}$$

Then by subadditivity,

$$m(E \setminus F) \le \sum_{n=0}^{\infty} m(E \setminus F_n) < \epsilon$$
 (88)

Finally,  $f_n$  converges uniformly on F and  $f_n|_F = g_n|_F$ , so  $f_n$  are continuous on F. Since uniform limit of continuous functions is continuous, the limit f is continuous on F.

This is an argument for the interval, but this can be generalized to more general sets.

## 4 Lebesgue Integration

Remember that Riemann integration is characterized by the approximation of step functions, which are the "building blocks" of Riemann integrable functions. To define the Lebesgue integral, we will start off with simple functions—which are a generalization of step functions. A function will be Lebesgue integrable if it can be approximated by these simple functions in some appropriate way. There are parallels between how we construct the Riemann and Lebesgue integral, namely that we can define the upper and lower integrals as the infimums and supremums of some set. However—while we were able to do this in "one shot" for the Riemann integral, the Lebesgue integral requires us to take intermediate steps. There are two ways that we can construct the Lebesgue integral.

- 1. We define the integral for simple functions  $\phi$  over finite measure. Since simple functions are made up of a linear combination of indicators, which are themselves bounded, the finite measure locks in the property that  $\int \phi$  will be bounded. This allows us to take the integral of simple functions that have real-valued coefficients—both positive and negative—but is inherently limiting as we can't define the integral over infinite measures. With this, we can define the integral of bounded measurable functions using the lower and upper integrals. This is similar to the construction of the Riemann integral. Furthermore, if we want to define the general integral, we have to now restrict what we have built up into the nonnegative case—a bit unnatural.
- 2. We first define the integral for nonnegative simple functions  $\phi$  over any measure. We are sacrificing the ability to integrate negative simple functions early on, but this doesn't matter since we will split functions into a positive and negative part anyways. The true advantage of doing this is that we gain the ability to define integrals for infinite measures. This allows us to define for positive (possibly unbounded) measurable functions, and finally when we define the general Lebesgue integral, we deal with the  $\infty \infty$  problem by defining it only if at least one of the positive or negative integrals are finite.

I personally think the second way is superior, since in the end we can define integrals over infinite measure. Furthermore, since we are only working with positive functions, we only need to define the lower integral rather than checking to see if the lower and upper integrals coincide.

As we redefine these integrals over and over (sort of like method overriding), we really want to preserve three properties of the integral.

1. Linearity.

$$\int_{E} (\alpha f + \beta g) = \alpha \int_{E} f + \beta \int_{E} g \tag{89}$$

2. Monotonicity

$$f \le g \implies \int_E f \le \int_E g$$
 (90)

3. Additivity.

$$A, B$$
 disjoint, measurable, and  $E = A \cup B \implies \int_{E} f = \int_{A} f + \int_{B} f$  (91)

## 4.1 Simple Functions

We will use  $\phi, \varphi, \psi$  to denote simple functions.

Definition 4.1 (Lebesgue Integral of Simple Functions over Finite Measure)

Suppose 
$$\phi = \sum_{k=1}^{n} a_k \chi_{E_k} : E \subset \mathbb{R} \to \mathbb{R}$$
 with  $a_k \in \mathbb{R}$ .

1. If  $m(E) < +\infty$ , then we can define the integral which is guaranteed to be a finite number.

$$\int_{E} \phi := \sum_{k=1}^{n} a_k m(E_k) \tag{92}$$

2. If  $a_k \geq 0$ , we define the integral which lives in  $\mathbb{R} \cup \{+\infty\}$ .

$$\int_{E} \phi := \sum_{k=1}^{n} a_k m(E_k) \tag{93}$$

This is well defined for any representation of  $\phi^b$ 

<sup>a</sup>As we have stated before, we could also define the Lebesgue integral of simple functions by letting  $a_k$  takes values in  $\mathbb{R}$ . But then, we might have a case where  $E=A\sqcup B$ , with  $m(A)=+\infty, m(B)=+\infty$ , and  $\phi=\chi_A-2\chi_B\Longrightarrow \int_E\phi=\infty-\infty$ . To prevent this from happening, some authors add the assumption that  $m(E)<+\infty$ , and I cover this case to make it as comprehensive as possible.

<sup>b</sup>We need this since the coefficients need not be unique. For example, we can write  $1 \cdot \chi_{[0,1]} + 1 \cdot \chi_{[0.5,1]} = 1 \cdot \chi_{[0,0.5]} + 2 \cdot \chi_{[0.5,1]}$ . If the  $E_i$ 's are disjoint, then this decomposition is unique and is called the **standard representation** of  $\phi$ .

#### Proof.

It is clear that the two definitions coincide if  $m(E) < +\infty$  and  $a_k \ge 0$  is true; it is the same formula.

For bounded functions, we will temporarily focus on the first definition, and for general functions, we rely on the second definition.

## Theorem 4.1 (Integral Properties for Simple Functions over Finite Measure)

Suppose  $\phi, \psi: E \subset \mathbb{R} \to \mathbb{R}$  are simple with  $m(E) < +\infty$ . Then, the following properties hold.

1. Linearity. For  $\alpha, \beta \in \mathbb{R}$ ,

$$\int_{E} (\alpha \phi + \beta \psi) = \alpha \int_{E} \phi + \beta \int_{E} \psi \tag{94}$$

2. Monotonicity.

$$\phi \le \psi \implies \int_{E} \phi \le \int_{E} \psi$$
 (95)

3. Additivity. If A, B are disjoint, measurable, and  $E = A \cup B$ , then

$$\int_{E} \phi = \int_{A} \phi + \int_{B} \phi \tag{96}$$

## Proof.

Listed.

- 1. We can subdivide sets s.t.  $\phi$  and  $\psi$  can be rewritten using the same finite family of sets  $A_k$ .
- 2. We can use (1) to rewrite

$$\int_{E} \psi - \int_{E} \phi = \int \underbrace{(\psi - \phi)}_{>0 \forall x} \ge 0 \tag{97}$$

3. Trivial by definition of simple functions.

## Example 4.1 (Step Function as Simple Function)

For  $a, b \in \mathbb{R}$ , with a < b, let  $f : [a, b] \longrightarrow \mathbb{R}$  be a step function. That is, there exists a partition  $a = x_0 < x_1 < \ldots < x_n = b$  and constants  $c_1, c_2, \ldots, c_n \in \mathbb{R}$  s.t.  $f(x) = c_i$  for all  $x \in (x_{i-1}, x_i)$  and each  $i = 1, \ldots, n$ . Then, f is equal to the following simple function, taken over all open intervals and the points  $x_j$  at the boundary of each interval.

$$f = \sum_{i=1}^{n} c_i \chi_{(x_{i-1}, x_i)} + \sum_{j=0}^{n} f(x_j) \chi_{\{x_j\}}$$
(98)

If we ignore the behavior of f on the partition points  $x_j$ 's, then f agrees almost everywhere with the simple function

$$\sum_{i=1}^{n} c_i \chi_{(x_{i-1}, x_i)} \tag{99}$$

## 4.2 Bounded Functions

This isn't the most popular way to define Lebesgue integrability, but I'd like to compare it to Riemann integral. First, note that a Riemann integral

## Definition 4.2 (Lower, Upper Lebesgue Integral)

Let  $f: E \subset \mathbb{R} \to \mathbb{R}$  be bounded, with  $m(E) < +\infty$ . Then, the **upper and lower Lebesgue integrals** are defined

$$\overline{L}f := \inf_{\phi} \left\{ \int \phi \mid f \le \phi, \phi \text{ simple} \right\}, \qquad \underline{L}f := \sup_{\phi} \left\{ \int \phi \mid \phi \le f, \phi \text{ simple} \right\}$$
 (100)

If the upper and lower Lebesgue integrals are equal, then f is said to be **Lebesgue integrable**.

This is exactly the same form as that of Riemann integration, notably that f must be bounded and we define integrability as the equivalence of the upper and lower integrals. The fact that m(E) is finite is realized implicitly with partitions, though in this sense it's more similar to the Riemann-Stieltjes integral. From this, it's pretty easy to intuit that the Lebesgue integral agrees with the Riemann integral for step functions. Let  $c_1, \ldots, c_n \in [0, \infty)$  and  $a = x_0 < x_1 < \ldots < x_n = b$  be a partition. Let  $f: [a, b] \longrightarrow [0, \infty]$  be a step function taking the value  $c_i$  on the interval  $(x_{i-1}, x_i)$  for  $i = 1, \ldots, n$ . Then the Riemann integral of f is simply

$$\int f(x) dx = \sum_{i=1}^{n} c_k |x_i - x_{i-1}|$$
(101)

The Lebesgue integral is

$$\int f d\mu = \sum_{i=1}^{n} c_i \mu((x_{i-1}, x_i)) + \sum_{j=0}^{n} f(x_j) \mu(\{x_j\}) = \sum_{i=1}^{n} c_k |x_i - x_{i-1}|$$
(102)

which agrees with the Riemann integral. In the Riemann integral, we write dx to indicate the variable that is being integrated over, but in the Lebesgue integral, we write  $d\mu$ , the measure which we are integrating over. We make this more rigorous in the following theorem.

## Theorem 4.2 (Riemann Integrability Implies Lebesgue Integrability)

If f is Riemann integrable, then it is Lebesgue integrable.

#### Proof.

Recall that f is Riemann integrable if

$$\sup_{P} L(P, f) = \inf_{P} U(P, f) \tag{103}$$

for a partition P. But for any L(P, f) (or U(P, f)), there exist simple  $\phi$  (or  $\psi$ ) s.t.

$$\int_{E} \phi = L(P, f), \qquad \left(\int_{E} \psi = U(P, f)\right) \tag{104}$$

So,

$$\sup_{P} L(P, f) \le \underline{L}f \le \overline{L}f \le \inf_{P} U(P, f) \tag{105}$$

where the first and third inequalities we just showed, and the middle inequality  $\underline{L}f \leq \overline{L}f$  comes from monotonicity. So if the inf = sup, then  $\underline{L}f = \overline{L}f$  has nowhere to go.

However, the converse is not true.

## Example 4.2 (Lebesgue Integrable but Not Riemann Integrable)

Consider the simple function (consisting of one characteristic function)  $\chi_{\mathbb{Q}\cap[0,1]}$ .  $\mathbb{Q}\cap[0,1]$  is a Lebesgue measurable set of  $\mathbb{R}$ , and we have  $\chi_{\mathbb{Q}\cap[0,1]} \geq 0$ , so its Lebesgue integral is given by the above definition:

$$\int_{\mathbb{R}} \chi_{\mathbb{Q} \cap [0,1]} d\lambda = 1 \cdot \lambda(\mathbb{Q} \cap [0,1]) = 0$$
(106)

Remember that continuous functions are Riemann integrable. There is indeed an analogous result between measurable functions and Lebesgue-integrable functions.

#### Theorem 4.3 (Measurable Functions are Lebesgue Integrable)

Let  $f: E \subset \mathbb{R} \to \mathbb{R}$  be measurable, bounded with  $m(E) < +\infty$ . Then, f is Lebesgue integrable.

## Proof.

We prove that  $\forall n, \exists$  a simple  $\phi_n, \psi_n$  s.t.  $\phi_n \leq f \leq \psi_n$  and  $\psi_n - \phi_n < 1/n$ . Then,

$$\overline{L}f - \underline{L}f \le \int \psi_n - \int \phi_n \le \frac{1}{n}m(E) \tag{107}$$

So take  $n \to \infty$ .

Now, we have a much larger class of functions we can integrate. It turns out that there is an iff condition for Lebesgue integrability.

## Theorem 4.4 (Lebesgue Integrable iff Set of Discontinuities have Measure 0)

A function  $f: E \subset \mathbb{R}$  is Lebesgue integrable iff f is continuous almost everywhere.

#### Proof.

To prove additivity, we'll need a specific lemma.

## Lemma 4.1 ()

Let  $f: E \subset \mathbb{R} \to \mathbb{R}$  be measurable, bounded with  $m(E) < +\infty$ , and  $A \subset E$  measurable. Then,

$$\int_{E} f \cdot \chi_{A} = \int_{A} f \tag{108}$$

#### Proof.

# Theorem 4.5 (Integral Properties for Bounded Measurable Functions over Finite Measure)

Suppose  $f, g : E \subset \mathbb{R} \to \mathbb{R}$  are bounded and measurable with  $m(E) < +\infty$ . Then, the following properties hold.

1. Linearity. For  $\alpha, \beta \in \mathbb{R}$ ,

$$\int_{E} (\alpha f + \beta g) = \alpha \int_{E} f + \beta \int_{E} g \tag{109}$$

2. Monotonicity.

$$f \le g \implies \int_{E} f \le \int_{E} g$$
 (110)

3. Additivity. If  $A \subset E$  is measurable  $B = A \setminus E$ , then

$$\int_{E} f = \int_{A} f + \int_{B} f \tag{111}$$

## Proof.

Listed.

1. For scalar multiplication, we just use  $\alpha\phi$ ,  $\alpha\psi$  in the simple function approximation. Now for the sums of functions, we can see that

$$f \le \psi_1, g \le \psi_2 \implies f + g \le \psi_1 + \psi_2 \implies \int (f + g) \le \int f + \int g$$
 (112)

$$f \ge \psi_1, g \ge \psi_2 \implies f + g \ge \psi_1 + \psi_2 \implies \int (f + g) \ge \int f + \int g$$
 (113)

(114)

and by taking the limit as the simple functions approach f,g, we get equality.

2

3. We can define  $f_1 = f \cdot \chi_A$ ,  $f_2 = f \cdot \chi_B$ .  $f_1$ ,  $f_2$  are measurable and bounded (as the product of measurable functions) implying that

$$\int_{E} f = \int_{E} (f_1 + f_2) = \int_{E} f_1 + \int_{E} f_2 = \int_{E} f \cdot \chi_A + \int_{E} f \cdot \chi_B = \int_{A} f + \int_{B} f$$
 (115)

where the final equality follows from the lemma.

## Corollary 4.1 ()

Suppose  $f: E \subset \mathbb{R} \to \mathbb{R}$  is bounded and measurable. Then

$$\left| \int_{E} f \right| \le \int_{E} |f| \tag{116}$$

#### Proof.

We know  $-|f| \le f \le |f|$ . By monotonicity, we have

$$\int -|f| \le \int f \le \int |f| \tag{117}$$

The next theorem is sort of like transferring convergence of functions to that of integrals. This results shouldn't be surprisingly since uniform convergence is so strong.

## Theorem 4.6 (Uniform Convergence Implies Convergence of Integrals)

Suppose  $f_n$  are bounded, measurable, and  $f_n \to f$  uniformly on E with  $m(E) < +\infty$ . Then,

$$\int_{E} f_n \to \int_{E} f \tag{118}$$

## Proof.

Since f is bounded, there exists  $N \in \mathbb{N}$  s.t.  $||f_N - f||_{\infty} \le 1$ . Then, by reverse triangle inequality (?),  $||f||_{\infty} \le ||f_N||_{\infty} + 1$ . Also, f is measurable as the limit of  $f_N$ . Given  $\epsilon > 0$ , find  $N_1$  s.t.  $\forall n \ge N_1$ ,

$$||f_n - f||_{\infty} \le \frac{\epsilon}{m(E)} \implies \left| \int f - \int f_n \right| = \left| \int_E (f - f_n) \right| \le \int_E |f - f_n| \le \frac{\epsilon}{m(E)} \cdot m(E) = \epsilon \quad (119)$$

Naturally we might see if this results holds with weaker assumptions. Unfortunately, this is not true for pointwise convergence.

#### Example 4.3 (Integrals Don't Converge Under Pointwise Convergence)

Let 
$$f_n(x) = n \cdot \chi_{[0,1/n]}(x)$$
, so  $f_n(x) \to 0$  a.e. in  $[0,1]$  but  $\int f_n(x) = 1$  for all  $n \in \mathbb{N}$ .

But not all hope is lost. This is where the key theorems of measure theory comes in. The following theorem is more general, since uniform convergence implies uniformly bounded, but it isn't used in practice as much (according to Kiselev).

## Theorem 4.7 (Bounded Convergence Theorem)

Suppose  $f_n$  are measurable, uniformly bounded.<sup>a</sup> Suppose  $f_n \to f$  pointwise on E, with  $m(E) < \infty$ . Then,

$$\int_{E} f_n \to \int_{E} f \tag{120}$$

 $a||f_n||_{\infty} \leq M < \infty$ 

#### Proof.

f is bounded because  $f_n$  are uniformly bounded, and it is measurable since it's a limit of measurable functions. Fix  $\epsilon > 0$ . by Egoroff,  $\exists F \subset E$  s.t.  $f_n \to f$  uniformly on F, and  $m(E \setminus F) \leq \frac{\epsilon}{4M}$ . Then,

$$\left| \int_{E} f - \int_{E} f_{n} \right| \le \int_{E} |f - f_{n}| = \int_{F} |f - f_{n}| + \int_{E \setminus F} |f - f_{n}| \tag{121}$$

For the first term,  $\exists N \in \mathbb{N}$  s.t. if  $n \geq N$ , then  $|f_n(x) - f(x)| \leq \frac{\epsilon}{2m(F)} \leq \epsilon 2$  for all  $x \in F$ . For the second term, we can bound this by  $2M \cdot \frac{\epsilon}{4M} \leq \frac{\epsilon}{2}$ . Therefore, the whole expression  $\leq \frac{\epsilon}{2} + \frac{\epsilon}{2} = \epsilon$ .

However, this assumption is still too strong for convergence and is therefore not used much in practice unlike other theorems (e.g. monotone convergence theorem and dominated convergence theorem). One nice part is that it works for unbounded functions, which is good since working with bounded functions is not the most natural assumption to have in practice. This is why we will step back and reconstruct the Lebesgue integral using positive—and possibly unbounded—simple functions (1st definition).

## 4.3 Positive Functions

If the  $A_i$ 's are just intervals in  $\mathbb{R}$ , then  $\phi$  reduces to a step function. But the entire problem with intervals is that they are too coarse. We can't work with them, so we generalize them to all measurable sets in  $(X, \mathcal{A})$ . The Riemann integral is built on an approximation scheme of a function, which we usually want to be continuous to satisfy this approximation, and so, if we want to build an approximation scheme for Lebesgue integrals, we want a similar scheme, i.e. if we take a sequence of simple measurable functions, I can get arbitrarily close to any measurable function f. This is exactly what we show below.

## Theorem 4.8 ()

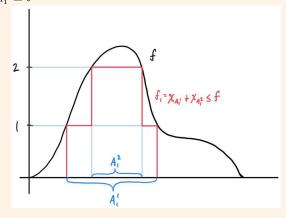
If  $f:(X,\mathcal{A})\longrightarrow [0,\infty]$  is measurable, there are simple measurable functions  $f_k:(X,\mathcal{A})\longrightarrow [0,\infty)$  s.t.

$$f_k \le f_{k+1} \text{ and } f = \lim_{k \to \infty} f_k$$
 (122)

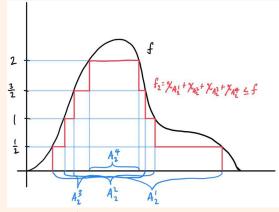
where the inequalities and limits are pointwise.

#### Proof.

We give a general picture of this proof for a function  $f: \mathbb{R} \longrightarrow [0, \infty]$ . We can first divide the codomain of the graph below into segments of t = 1, 2, ..., and take the preimage of all these units under f to get  $f_1$ . More specifically,  $A_1^t = f^{-1}([t, \infty])$  for all t. By measurability of f,  $A_1^t$  is measurable, and we can assign  $f_1 = \chi_{A_1^1} + \chi_{A_1^2} \leq f$ .



Doing this again with finer subintervals of the codomain gives us, with  $f_2 = \chi_{A_2^1} + \chi_{A_2^2} + \chi_{A_2^3} + \chi_{A_2^4} \leq f$ .



and in general, we have  $f_k = \sum_{j=1}^{\infty} \frac{1}{2^{k-1}} \chi_{A_k^j}$ . But we said a simple function is a *finite* sum, and if  $\infty$  is in the range of f, then this becomes a problem. We can quickly fix this by just truncating the summation at a certain point in the codomain ( $f_1$  only considers intervals up to 1,  $f_2$  up to 2 and so on), ultimately giving us

$$f_k = \sum_{j=1}^{k2^{k-1}} \frac{1}{2^{k-1}} \chi_{A_k^j} \tag{123}$$

Finally, we can learn how to integrate. We require the positiveness condition on f below because our previous theorem on approximating arbitrary functions with simple measurable functions  $f_k$  requires that it be positive, too.

## 4.4 Positive Measurable Functions

The next natural step to generalize Lebesgue integral is to look at unbounded functions and/or infinite measures. In order to do this, we must start off by stepping back into only nonnegative functions first.

Unlike Riemann integration and Lebesgue integration of signed bounded functions, which looks at both the supremum and infimum of integrals of simple functions, Lebesgue integration of positive only looks at the supremum, given that f is nonnegative, so for all these f, the Lebesgue integral always exists.

## Definition 4.3 (Lebesgue Integral on Positive Measurable Functions)

Let  $f: E \subset \mathbb{R} \to \mathbb{R}$  satisfy  $f \geq 0$ , with E measurable (but not necessarily that  $m(E) < +\infty$ ). Then, the **Lebesgue integral** is defined

$$\int_{E} f := \sup \left\{ \int_{E} h \mid 0 \le h \le f, h \text{ measurable, bounded} \right\}$$
 (124)

$$= \sup \left\{ \int_{E} \phi \mid 0 \le \phi \le f, \phi \text{ simple} \right\}$$
 (125)

#### Proof.

The only thing to show is that it suffices to check for simple functions only, so all we have to do is check the latter form.

## Lemma 4.2 (Chebyshev)

Suppose  $f: E \subset \mathbb{R} \to \mathbb{R}$ ,  $f \geq 0$ , and f is measurable. Then for all  $\alpha > 0$ ,

$$m(\lbrace x \in E \mid f(x) \ge \alpha \rbrace) \le \frac{1}{\alpha} \int_{E} f \tag{126}$$

<sup>a</sup>In probability, this refers to a bound on the variance of a random variable, but in analysis, it seems to be a more general result of the bounds of form  $\int |f|^p$ .

#### Proof.

Let us call the set on the LHS  $E_{\alpha}$ . Then,  $E_{\alpha}$  is measurable and define  $\phi(x) = \alpha \chi_{E_{\alpha}}(x)$ . Then,  $\phi(x) \leq f(x)$ , and so

$$\int f \ge \int \phi = \alpha m(E_{\alpha}) \tag{127}$$

## Theorem 4.9 (Vanishing Integral iff a.e. Vanishing Function)

Suppose  $f \geq 0$  on E. Then,

$$\int_{E} f = 0 \iff f = 0 \text{ a.e. on } E$$
 (128)

#### Proof.

We prove bidirectionally.

1.  $(\rightarrow)$ . By Chebyshev, we see that

$$m(\underbrace{\{x \in E \mid f(x) \ge 1/n\}}_{E}) \le n \int f = 0 \tag{129}$$

for all  $n \in \mathbb{N}$ . But  $\{x \in E \mid f(x) > 0\} = \bigcup_{n=1}^{\infty} E_n$ . By countable additivity,  $m(\{x \in E \mid f(x) > 0\}) = 0$ .

2. ( $\leftarrow$ ). Since f = 0 a.e., any  $0 \le \phi \le f - \phi$  simple—will satisfy  $m(\{x \in E \mid \phi(x) = 0\}) = 0$ , but it will never get off 0. Therefore,  $\int_E \phi = 0$ .

## Theorem 4.10 (Integral Properties for Nonnegative Measurable Functions)

Suppose nonnegative  $f, g : E \subset \mathbb{R} \to \mathbb{R}$  are measurable. Then,

1. Linearity. For all  $\alpha, \beta \geq 0$ ,

$$\int_{E} (\alpha f + \beta g) = \alpha \int f + \beta \int g \tag{130}$$

2. Monotonicity.

$$f \le g \implies \int f \le \int g$$
 (131)

3. Additivity. If A, B are disjoint with  $E = A \cup B$ , then

$$\int_{E} f = \int_{A} f + \int_{B} f \tag{132}$$

<sup>&</sup>lt;sup>a</sup>Note that we have  $\geq 0$  since we are dealing with positive functions!

#### Proof.

Listed.

- 1. We can easily show  $\int \alpha f = \alpha \int f$  simply by multiplying the simple functions  $\phi$  by  $\alpha$ . For integrals of sums of functions, we show the following.
  - (a)  $\int f + \int g \leq \int (f+g)$ . Since given simple  $\phi_1, \phi_2$  with  $\phi_1 \leq f, \phi_2 \leq g$ , we have  $\phi_1 + \phi_2 \leq f + g$ , and so by taking the supremum, we have the bound.
  - (b)  $\int f + \int g \ge \int (f+g)$ . Suppose h is simple s.t.  $0 \le h \le f+g$ . Define<sup>a</sup>

$$\ell \coloneqq \min\{h, f\}, \qquad k \coloneqq h - \ell \tag{133}$$

Note that  $\ell, k$  are both measurable. Furthermore,  $\ell$  is at most h (which is simple) and so is bounded, and k is also bounded since it is bounded h minus some nonnegative  $\ell$  that is at most h. Therefore, we invoke our previous definition of the Lebesgue integral for bounded functions to get

$$\int k + \int \ell = \int h \tag{134}$$

Since  $\ell \leq f$  and  $k = h - \ell \leq g$ , by monotonicity we have  $\int k \leq \int g$  and  $\int \ell \leq \int f$ . Substituting this in gives

$$\int h \le \int f + \int g \implies \int (f+g) = \sup_{h} \int h \le \int f + \int g \tag{135}$$

2. Direct.

3.

We know that the integral doesn't behave well under weak kinds of convergence, e.g. pointwise or a.e. The following theorem gives us a slightly stronger assumption.

#### Theorem 4.11 (Monotone Convergene Theorem (MCT))

Given a nondecreasing sequence of nonnegative measurable functions  $f_1 \leq f_2 \leq f_3 \leq \ldots : X \longrightarrow [0, \infty]$ , its limit  $\lim_{k \to \infty} f_k$  always exists (since  $f_k$  is nondecreasing), is measurable, and

$$\int \lim_{k \to \infty} f_k \, d\mu = \lim_{k \to \infty} \int f_k \, d\mu \tag{136}$$

## Proof.

Suppose  $\lim_{n\to\infty} \int f_n = \alpha$  (could be infinity),  $f(x) \geq f_n(x)$  for all x, n, so  $\int f \geq \alpha$ . Consider  $0 \leq \phi \leq f$ ,  $\phi$  simple. Take  $0 < c < 1^a$  and define

$$E_n\{x \mid f_n(x) \ge c\phi(x)\}\tag{137}$$

Then, the  $E_n$  are increasing and  $\bigcup_{n=1}^{\infty} E_n = E^b$ . Observe that

$$\int_{E_n} f_n \ge c \int_{E_n} \phi \tag{138}$$

Suppose that  $\phi(x) = \sum_{k=1}^{M} a_k \chi_{F_k}(x)$ . Then,

$$\int_{E_n} f = \sum_{k=1}^M a_k m(F_k \cap E_n) \to \sum_{k=1}^M a_k m(F_k) = \int_E \phi \text{ as } n \to +\infty$$
 (139)

<sup>&</sup>lt;sup>a</sup>Intuitively, we are trying to split the f+g into  $\ell$  and k, such that  $\ell+k=h\leq f+g$ .

Therefore,

$$\lim_{n \to \infty} \int_{E} f_n \ge c \int_{E} \phi \forall c < 1 \implies \text{also true for } c = 1$$
 (140)

Since  $\phi \leq f$  is arbitrary, we get  $\lim_{n\to\infty} \int f_n \geq \int f$ .

<sup>a</sup>We introduce the c because we can then claim that the union of  $E_n$  is E.

This allows us to integrate the limit of nice functions  $f_k$  by integrating these  $f_k$  first and then finding what the values converge to. The huge problem with Riemann integrals is that this theorem doesn't hold, but it is the case for Lebesgue integration. In practice, the MCT is not as useful, but the following is.

## Lemma 4.3 (Fatau's Lemma)

Suppose  $f_n \geq 0$  and it converges to f. Then,

$$\int_{E} \liminf f_n \le \liminf \int_{E} f_n \tag{141}$$

#### Proof.

Define  $g_k(x) = \int_{j \ge k} f_j(x)$ . Note that  $g_k(x) \le f_k(x)$  by definition, and  $g_k(x)$  is increasing. Since by definition  $\lim_{k \to \infty} g_k(x) = \liminf_{k \to \infty} f_k(x)$ , by MCT,

$$\lim_{k \to \infty} \int_{E} g_k = \int \liminf f_k \tag{142}$$

But since  $\int g_k \leq \int f_k$  for all k,

$$\liminf \int f_k \ge \int \liminf f_k \tag{143}$$

Note that equality does not have to hold. For example, take  $f_n(x) = \chi_{[n,n+1]}(x)$ .

## Definition 4.4 (Lebesgue Integrability)

A positive measurable function f is **Lebesgue integrable** if  $\int f < +\infty$ .

## Theorem 4.12 ()

If f is integrable on E, then  $f(x) < +\infty$  a.e. on E.

## Proof.

By Chebyshev,

$$m(\{x \in E \mid f(x) \ge n\}) \le \frac{1}{n} \int_{E} f$$
 (144)

As  $n \to \infty$ , the LHD is bounded above by all  $\epsilon > 0$ .

$$m(\lbrace x \in E \mid f(x) = \infty \rbrace) = m\left(\bigcap_{n=1}^{\infty} E_n\right)$$
(145)

Here is a variant of MCT.

<sup>&</sup>lt;sup>b</sup>This may not be true is c = 1.

## Lemma 4.4 (Levi's Lemma)

Suppose  $f_n \geq 0$  is increasing and  $\int_E f_n \leq M$  for all n. Then,

- 1.  $f(x) = \lim_{n \to \infty} f_n(x)$  is integrable, and
- 2.  $\int f \leq M$ .

#### Proof.

By MCT,  $\int_E f_n \to \int_E f$ .

## 4.5 Lebesgue Integral for Signed Functions

## Definition 4.5 (Lebesgue Integral for Signed Functions)

Given  $f: E \subset \mathbb{R} \to \mathbb{R} \cup \{\pm \infty\}$ , let us write

$$f = f^+ - f^-, \qquad f^+ := \max\{0, f\}, f^- := \max\{-f, 0\}$$
 (146)

f is Lebesgue integrable iff  $f^+$  and  $f^-$  and we define the Lebesgue integral of f as

$$\int f := \int f^{+} - \int f^{-} \tag{147}$$

given that at least one of these integrals is finite. If one is infinite and the other is finite, then we can call it infinite. If we have *both* infinite integrals, then the integral doesn't exist.

If  $f^+, f^-$  are Lebesgue integrable, then it can only be  $\pm \infty$  on a set of measure 0, so it generally won't affect anything. Also, there is a difference between the *existence* of the integral and a function *being* Riemann-integrable.

Since  $|f| = f^+ + f^-$ , f is also Lebesgue integrable if

$$\int |f| \, d\mu < \infty \tag{148}$$

since by triangle inequality, we have

$$\left| \int f \, d\mu \right| \le \int |f| \, d\mu \tag{149}$$

## Theorem 4.13 ()

 $f^{\pm}$  integrable iff |f| integrable.

#### Proof.

 $|f| = f^+ + f^-.$ 

## Theorem 4.14 (Integral Properties of Signed Measurable Functions)

Suppose  $f, g: E \subset \mathbb{R} \to \mathbb{R}$  are measurable. Then,

1. Linearity. For all  $\alpha, \beta \in \mathbb{R}$ ,

$$\int_{E} (\alpha f + \beta g) = \alpha \int f + \beta \int g \tag{150}$$

2. Monotonicity.

$$f \le g \implies \int f \le \int g$$
 (151)

3. Additivity. If A, B are disjoint with  $E = A \cup B$ , then

$$\int_{E} f = \int_{A} f + \int_{B} f \tag{152}$$

## Proof.

Listed. As always, linearity is nontrivial.

- 1. For scalar multiplication, we divide into 2 cases.
  - (a)  $\alpha > 0$ . Then,  $(\alpha f)^+ = \alpha f^+$  and  $(\alpha f)^- = \alpha f^-$ , so

$$\int (\alpha f) = \int \alpha f^{+} - \int \alpha f^{-} = \alpha \int f \tag{153}$$

(b)  $\alpha < 0$ . Then,  $(\alpha f)^+ = -\alpha f^+$  and  $(\alpha f)^- = \alpha f^+$ , so

$$\int (\alpha f) = \int -\alpha f^{-} + \int \alpha f^{+} = \alpha \left( \int f^{+} - \int f^{-} \right) = \alpha \int f \tag{154}$$

2. For addition, note that  $|f+g| \leq |f| + |g|$ , so it is integrable. So

$$\int (f+g) = \int (f+g)^{+} - \int (f+g)^{-}$$
(155)

Now observe that

$$(f+g)^{+} - (f+g)^{-} = f+g = f^{+} + g^{+} - f^{-} - g^{-}$$
(156)

$$(f+g)^{+} + f^{-} + g^{-} = (f+g)^{-} + f^{+} + g^{+}$$
(157)

Since it is nonnegative, our previous properties of linearity holds, and then we can just rearrange.

3.

4.

## Definition 4.6 (Normed Vector Space of Lebesgue Integrable Functions)

The set of all functions  $f:(X,\mathcal{A},\mu)\longrightarrow\mathbb{R}$  that are Lebesgue integrable is denoted  $\mathcal{L}^1(X,\mathcal{A},\mu;\mathbb{R})$ , or for short  $\mathcal{L}^1(X,\mathcal{A},\mu)$ .

## Theorem 4.15 ()

Suppose  $f:(\mathbb{R},\mathcal{A},\mu)\longrightarrow\mathbb{R}$  is 0 almost everywhere. Then f is Lebesgue integrable with

$$\int_{\mathbb{R}} f \, d\mu = 0 \tag{158}$$

If  $g: \mathbb{R} \longrightarrow \mathbb{R}$  is such that  $f = g \mu$ -almost everywhere, then

$$\int_{\mathbb{R}} f \, d\mu = \int_{\mathbb{R}} g \, d\mu \tag{159}$$