Probability and Statistical Inference

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4 Lebesgue Measure

4.1 Motivation: From Probability to Length (Number Theory Version)

Now suppose that we need to draw a real number at random from the interval [0, 1]. What is the probability that it lies in [0, 0.47)?

In a discrete setting, we count favorable outcomes. Here, outcomes are uncountably infinite—so what does "probability" mean?

We can break down the problem by **picking digits** one by one.

- Pick a number: 0.xy...
- Either $x \in \{0, 1, 2, 3\}$ or $x = 4 \cap y \in \{0, 1, \dots, 0.6\}$.

possibilities = $(4 \cdot 10) + (1 \cdot 7) = 47$ out of $(10 \cdot 10)$ possibilities.

 $\mathbb{P}(w \in [0, 0.47]) = \frac{47}{100} = 0.47$ Some shape of Uniform CDF.

4.2 Defining Lebesgue Measure on Intervals

Definition 4.2.1 (Lebesgue Measure on [0,1]). Let $a,b \in [0,1]$ with a < b. The Lebesgue measure μ assigns:

$$\mu([a,b)) = b - a.$$

In a more mathematical sense, the "length" of an interval is a "size" to a subset of any general set. Therefore, the essence of a measure, or Lebesgue measure on \mathbb{R} , intends to describe the "size" of almost any general subset in \mathbb{R} by extending this rule to finite unions of disjoint intervals and then, with more care, to broader collections of subsets.

This seems natural — the length of an interval. But to treat it as a true measure in probability context, we need to verify that it satisfies the Kolmogorov axioms and additional key properties.

4.3 Properties of Lebesgue Measure on \mathbb{R}

Proposition 4.3.1 Basic Properties:

Let μ denote Lebesgue measure on \mathbb{R} :

- 1. **Length:** $\mu([a,b)) = b a$
- 2. Normalization: $\mu([0,1]) = 1$
- 3. Translation Invariance: $\mu(x+A) = \mu(A), \ \forall x \in \mathbb{R}$

Proposition 4.3.2 Special Cases:

- 1. Open intervals: $\mu((a,b)) = \mu([a,b]) = \mu([a,b]) = b a$
- 2. Single element set: $\mu(\{a\}) = 0$
- 3. Countable set: $\mu(\mathbb{Q} \cap [0,1]) = 0$

Remark 4.1 Lebesgue measure assigns zero measure to all finite or countable sets — even dense ones like $\mathbb{Q} \cap [0,1]$.

4.4 What's Missing?

Remark 4.2 Why Care About Measurability? So far, we've assigned lengths to intervals and even countable unions. But it turns out — surprisingly — that not all subsets of \mathbb{R} can be measured in this way, namely:

Example 4.4.1 Banach—Tarski Paradox (Informal Version) In 3-dimensional space, there exists a decomposition of a solid ball into a finite number of disjoint pieces, which can be reassembled — using only rotations and translations — into two identical copies of the original ball.

This construction:

- Uses only finitely many pieces.
- Does not rely on scaling or stretching.
- Crucially depends on the Axiom of Choice.

But it cannot be carried out with measurable pieces.

Implication: Any measure that is translation invariant and countably additive cannot be defined on all subsets of \mathbb{R}^3 if we hold onto the axiom of choice.

Remark 4.3 To avoid these paradoxes, we restrict our attention to a special collection of subsets called measurable sets. In the next chapter, we will construct Lebesgue measure more carefully, using the concept of outer measure and the Carathéodory criterion to define which sets can be measured consistently.