Lecture 3: Aug 22

Last time

• Set theory (1.1)

Today

- Set theory (1.1)
- Axiomatic Foundations (1.2)

Theorem For any three events, A, B, and C, defined on a sample space S,

1. Commutativity

$$A \cup B = B \cup A$$
,
 $A \cap B = B \cap A$;

2. Associativity

$$A \cup (B \cup C) = (A \cup B) \cup C,$$

$$A \cap (B \cap C) = (A \cap B) \cap C;$$

3. Distributive Laws

$$A \cap (B \cup C) = (A \cap B) \cup (A \cap C),$$

$$A \cup (B \cap C) = (A \cup B) \cap (A \cup C);$$

4. DeMorgan's Laws

$$(A \cup B)^c = A^c \cap B^c,$$

$$(A \cap B)^c = A^c \cup B^c;$$

We show the proof of $A \cap (B \cup C) = (A \cap B) \cup (A \cap C)$ in the distributive laws. Caution: Venn diagrams are helpful in visualization, but they do not constitute a formal proof. To prove that two sets are equal, we need to show that each set contains the other. *proof:*

- $A \cap (B \cup C) \subset (A \cap B) \cup (A \cap C)$: Let $x \in (A \cap (B \cup C))$. By definition of intersection, $x \in (B \cup C)$ that is, either $x \in B$ or $x \in C$. Since x also must be in A, we have that either $x \in (A \cap B)$ or $x \in (A \cap C)$; therefore, $x \in ((A \cap B) \cup (A \cap C))$.
- $(A \cap B) \cup (A \cap C) \subset A \cap (B \cup C)$: Let $x \in ((A \cap B) \cup (A \cap C))$. This implies that $x \in (A \cap B)$ or $x \in (A \cap C)$. If $x \in (A \cap B)$, then x is in both A and B. Since $x \in B$, then $x \in (B \cup C)$ and thus $x \in (A \cap (B \cup C))$. It follows the same argument when $x \in (A \cap C)$, we still have $x \in (A \cap (B \cup C))$.

Definition Two events A and B are disjoint (or mutually exclusive) if $A \cap B = \emptyset$. The events A_1, A_2, \ldots are pairwise disjoint (or mutually exclusive) if $A_i \cap A_j = \emptyset$ for all $i \neq j$.

Definition If A_1, A_2, \ldots are pairwise disjoint and $\bigcup_{i=1}^{\infty} A_i = A_1 \cup A_2 \cup \cdots = S$, then the collection of A_1, A_2, \ldots forms a partition of S.

Example The sets $A_i = [i, i+1), i = 0, 1, 2, ...$ form a partition of $[0, \infty)$.

Basics of Probability Theory

When an experiment is performed, the realization of the experiment is an outcome in the sample space. If the experiment is performed a number of times, then

- different outcomes may occur each time
- some outcomes may repeat
- the "frequency of occurrence" of an outcome can be thought of as a probability

However, we **do not** define probabilities in terms of frequencies but instead take the mathematically simpler axiomatic approach. The axiomatic approach is not concerned with the interpretations of probabilities, but is concerned only that the probabilities are defined by a function satisfying the axioms. Interpretations of the probabilities are quite another matter:

- The "frequency of occurrence" of an event is one example of a particular interpretation of probability.
- Another possible interpretation is a subjective one, where we can think of the probability as a belief in the chance of an event occurring.

Axiomatic Foundations

For each event A in the sample space S, we want to associate with A a number between zero and one that will be called the probability of A, denoted by $\Pr(A)$. The domain of \Pr is the set where the arguments of the function $\Pr(\cdot)$ are defined. It is natural to define the domain of \Pr as all subsets of S, that is for each $A \subset S$, we define $\Pr(A)$ as the probability that A occurs. However, there are some technical difficulties to overcome which requires us to familiarize with the following.

Definition A collection of subsets of S is called a *sigma algebra* (or *Borel field*), denoted by \mathcal{B} , if it satisfies the following three properties:

- 1. $\emptyset \in \mathcal{B}$ (the empty set is an element of \mathcal{B}).
- 2. If $A \in \mathcal{B}$, then $A^c \in \mathcal{B}$ (\mathcal{B} is closed under complementation).
- 3. If $A_1, A_2, \dots \in \mathcal{B}$, then $\bigcup_{i=1}^{\infty} A_i \in \mathcal{B}$ (\mathcal{B} is closed under countable unions).

From Property (1) and (2), we see that the empty set and its complement S (since $S = \emptyset^c$) are always in a sigma algebra. In fact, they construct the *trivial* algebra $\{\emptyset, S\}$ which is the smallest sigma algebra.

By DeMorgan's Law, (3) can be replaced by:

3'. if
$$A_1, A_2, \dots \in \mathcal{B}$$
, then $\bigcap_{i=1}^{\infty} A_i \in \mathcal{B}$.

This is because:

Example If S is finite or countable (where the elements of S can be put into one to one correspondence with a subset of the integers), then these technicalities really do not arise, for we define for a given sample space S,

$$\mathcal{B} = \{\text{all subsets of } S, \text{ including } S \text{ itself}\}.$$

If S has n elements, there are 2^n sets in \mathcal{B} (why?).[hint: for each element, it is either in or out of a subset, so 2 choices].

Example Let $S = (-\infty, \infty)$, be the real line. Then \mathcal{B} is chosen to contain all sets of the form

$$[a, b], (a, b], (a, b), \text{ and } [a, b)$$

for all real numbers a and b. Also, from the properties of \mathcal{B} , it follows that \mathcal{B} contains all sets that can be formed by taking (possibly countably infinite) unions and intersections of sets of the above varieties.

We now define a probability function.

Definition Given a sample space S and an associated sigma algebra \mathcal{B} , a probability function is a function Pr with domain \mathcal{B} that satisfies

- 1. $Pr(A) \ge 0$ for all $A \in \mathcal{B}$.
- 2. Pr(S) = 1.
- 3. If $A_1, A_2, \dots \in \mathcal{B}$ are pairwise disjoint, then $\Pr(\bigcup_{i=1}^{\infty} A_i) = \sum_{i=1}^{\infty} \Pr(A_i)$.

The above three properties are usually referred to as the Axioms of Probability (or the Kolmogorov Axioms, after A. Kolmogorov, one of the fathers of probability theory). Any function that satisfies the Axioms of Probability is called a probability function.

Example Consider the simple experiment of tossing a fair coin (just once), so $S = \{H, T\}$. A reasonable probability function is the one that assigns equal probabilities to heads and tails, that is,

$$\Pr(\{H\}) = \Pr(\{T\}).$$

Since $S = \{H\} \cup \{T\}$, we have, from Axiom 2, $\Pr(\{H\} \cup \{T\}) = 1$. Also, $\{H\}$ and $\{T\}$ are disjoint, so $\Pr(\{H\} \cup \{T\}) = \Pr(\{H\}) + \Pr(\{T\})$. Collectively, we have

$$\Pr(\{H\}) = \Pr(\{T\})$$

 $\Pr(\{H\} \cup \{T\}) = 1$
 $\Pr(\{H\} \cup \{T\}) = \Pr(\{H\}) + \Pr(\{T\})$

Therefore, $Pr(\lbrace H \rbrace) = Pr(\lbrace T \rbrace) = \frac{1}{2}$.

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