STA2001 Probability and Statistics (I)

Lecture 1

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Question

What is probability theory and statistics?

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- Probability theory: a branch of mathematics concerned with the analysis of random phenomena, cf. Encyclopedia Britannica.
 - Random phenomena: flipping a coin, rolling a die, winning a lottery, stock index
 - Probability: the tool we use to analyze the random phenomena

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 - Random phenomena: flipping a coin, rolling a die, winning a lottery, stock index
 - Probability: the tool we use to analyze the random phenomena
- Statistics: the theory for the analysis of the <u>data</u>, how to extract <u>information</u> from data is the core of statistics, information can be used for making decisions and predictions
 - Data: observation/measurements of random phenomena
 - Information: data becomes information once it has been analyzed in some fashion, cf. Wikipedia.

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What is the importance of probability theory and statistics?

fundamental for many disciplines in science and engineering such as biology, machine learning, big data, artificial intelligence, signal processing, and many others!

A Question Throughout This Course

Question

Facing these random phenomena in our daily life, how would you build a mathematical framework to study them in a rigorous way?

In this course, we will review how mathematicians build probability theory to study random phenomena.

Section 1.1 Properties of Probability

Fundamental Concepts

Definition[Experiment]

Any procedure that can be infinitely repeated and has a well-defined set of possible outcomes.

Definition[Random Experiment]

An experiment is said to be random if it has more than one possible outcomes.

Definition[Sample Space]

Given a random experiment, the collection of all possible outcomes is called the sample space, denoted by S.

Fundamental Concepts

Definition[Event]

Given a sample space S, an event A is a set that contains part of outcomes in S; that is, $A \subseteq S$.

Definition[An event A has occurred]

When a random experiment is performed, if the outcome of the experiment is in A, then we say that the event A has occurred.

Example 1

Throwing a fair 6-sided die

- 1. This is a random experiment
- 2. Sample space $S=\{1,2,3,4,5,6\}$
- 3. Event $A = \{1,2\}$



- 4. Throw the die, if the outcome is either 1 or
 - 2, then A has occurred.

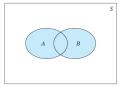
Set theory: fundamental role in probability theory

- Algebra [Reunion of broken parts]: the study of mathematical symbols and the rules for manipulating these symbols.
- Set: a collection of distinct elements
- Ø: the null or empty set

In the following, let A and B be two sets.

- ▶ $A \subseteq B$: A is a subset of B (every element of A is also an element of B).
- ▶ $A \cup B$: the union of A and B (set of elements that belong to either A or B).
- $ightharpoonup A \cap B$: the intersection of A and B.
- ► A': the complement of A in S is the set of all elements in S that are not in A.

- $ightharpoonup A_1, A_2, \cdots, A_k$ are said to be
 - 1. mutually exclusive if $A_i \cap A_j = \emptyset, i \neq j$
 - 2. exhaustive if $A_1 \cup A_2 \cup \cdots \cup A_k = S$
 - 3. mutually exclusive and exhaustive if 1 & 2 holds.

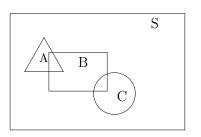


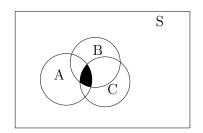
Commutative laws:

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

Associative Law:

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





Distributive law:

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

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

► De Morgan's law

$$(A \cup B)' = A' \cap B', \quad (A \cap B)' = A' \cup B'$$

Example 1 (Continued)

Recall

$$S = \{1, 2, 3, 4, 5, 6\}, \quad A = \{1, 2\}$$

Let

$$B = \{2, 3, 4\}, \quad C = \{5, 6\}$$

What is $A \cap B$, $A \cap (B \cup C)$?

An Intuitive Definition of Probability

Problem

how to define the probability of an event A, (the chance of A occurring)

An intuitive idea:

- repeat the experiment a number of times, say n times count the number of times that event A actually occurs, $\mathcal{N}(\mathsf{A})$
- $ightharpoonup rac{\mathcal{N}(A)}{n}$ is called the relative frequency of event A in n repetitions of the experiment

Example 1 (Continued)

$$S = \{1, 2, 3, 4, 5, 6\}, A = \{1, 2\}$$

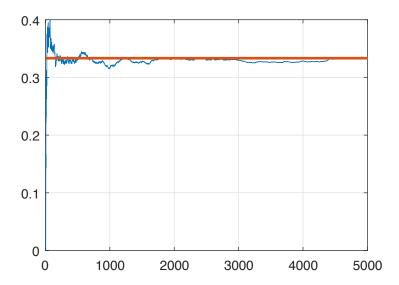
Outcome is either 1 or $2 \Rightarrow A$ occurs

Numerical simulation by computer programs shows

$$\frac{\mathcal{N}(A)}{n} \to \frac{1}{3}, \quad \text{as } n \to \infty$$

The number that $\frac{\mathcal{N}(A)}{n}$ goes to as $n \to \infty$ is called the probability of event A and is denoted by $P(A) = \lim_{n \to \infty} \frac{\mathcal{N}(A)}{n}$

Example 1 (Continued)



Definition of Probability (Probability Axioms)

Definition[Probability]

A real-valued, set function P that assigns to each event A in the sample space S, a number P(A), called the probability of the event A such that the following properties are satisfied:

- 1. $P(A) \geq 0$.
- 2. P(S) = 1.
- 3. if A_1, A_2, A_3, \cdots are countable and mutually exclusive events

$$P(A_1 \cup A_2 \cup \cdots) = P(A_1) + P(A_2) + \cdots$$

or equivalently,

$$P(\bigcup_{i=1}^{\infty}A_i)=\sum_{i=1}^{\infty}P(A_i)$$

Probability Axioms

The Kolmogorov axioms are the foundations of probability theory introduced by Andrey Kolmogorov in 1933.



Figure: Andrey Kolmogorov (25 April 1903 – 20 October 1987) was a Soviet mathematician who contributed to the mathematics of probability theory, topology, intuitionistic logic, turbulence, classical mechanics, algorithmic information theory and computational complexity.

Property 1: For each event A, P(A) = 1 - P(A').

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$$S = A \cup A', \quad A \cap A' = \emptyset$$

$$1 = P(S) = P(A \cup A') = P(A) + P(A') \Rightarrow P(A) = 1 - P(A')$$

Property 2: $P(\emptyset) = 0$. By property 1 and take A' = S.

Property 3: If events A and B are such that $A \subseteq B$, then

$$P(A) \leq P(B)$$

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$$P(A) \leq P(B)$$

$$B=B\cup A=(B\cup A)\cap S=(B\cup A)\cap (A'\cup A)=(B\cap A')\cup A$$
 note that $(B\cap A')\cap A=\emptyset$ and $P(B\cap A')\geq 0$

 $P(B) = P((B \cap A') \cup A) = P(B \cap A') + P(A) > P(A)$

Property 4: For each event A, $P(A) \le 1$.

$$P(S) = 1 = P(A \cup A') = P(A) + P(A') \ge P(A)$$

Property 5: For any two events A and B.

$$P(A \cup B) = P(A) + P(B) - P(A \cap B)$$

Property 4: For each event A, $P(A) \le 1$.

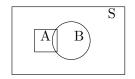
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Property 5: For any two events A and B.

$$P(A \cup B) = P(A) + P(B) - P(A \cap B)$$

$$A \cup B = (A \cup B) \cap S = (A \cup B) \cap (A \cup A') = A \cup (A' \cap B)$$

$$A \cup B = A \cup (A' \cap B)$$
, where $A \cap (A' \cap B) = \emptyset$



$$P(A \cup B) = P(A) + P(A' \cap B)$$
 (1)

$$B = B \cap S = B \cap (A \cup A')$$

$$B = (A \cap B) \cup (A' \cap B), \quad \text{where } (A \cap B) \cap (A' \cap B) = \emptyset$$

$$P(B) = P(A \cap B) + P(A' \cap B) \tag{2}$$

$$(1) + (2) \Rightarrow P(A \cup B) = P(A) + P(B) - P(A \cap B)$$

Probability Space*

A probability space is a triple (S, F, P)

- 1. S: the sample space
- 2. F is a σ -algebra on S, a collection of subsets of S, and called the event space

$$\int \bullet S \in F$$

- 3. $P: F \rightarrow [0,1]$ is the probability measure such that

$$P(A) \geq 0, \forall A \in F, \quad P(S) = 1, \quad P(\bigcup_{i=1}^{\infty} A_i) = \sum_{i=1}^{\infty} P(A_i)$$

for countable and mutually exclusive A_1, A_2, \cdots

Note: This slide is included here for your possible interest but not included in the exam. □ ► ◆□ ► ◆□ ► ◆□ ► □ ♥ Q ○ 32/32

Section 1.2 Method of Enumeration (Permutation and Combination)

Motivation

Why enumeration?

For some cases, to define and calculate P(A) can be converted to count the number of outcomes in $A \to$ counting techniques.

Assumption 1: S contains m possible outcomes

$$e_k$$
, $k = 1, 2, \dots, m$, i.e., $S = \{e_1, e_2, \dots, e_m\}$.

Assumption 2: The *m* outcomes are "equally likely"

$$P(\lbrace e_k \rbrace) = \frac{1}{m}, \quad k = 1, \cdots, m.$$

Extension of rolling die example.

$$S=\{1,2,3,4,5,6\}, P(\{k\})=\frac{1}{6}, k=1,\cdots,6.$$



Motivation

Then

$$P(A) = \frac{N(A)}{N(S)},$$

where N(X) is the number of outcomes in $X \subseteq S$.

- ▶ It can be verified P(A) is a well-defined probability function that satisfies the probability axioms.
- ▶ To calculate $P(A) \Leftrightarrow$ to count the number of elements in A and in S under Assumptions $1\&2 \Rightarrow$ links to the counting techniques, e.g., the method of enumeration.

Counting Techniques

Problem

To develop techniques for counting the number of outcomes associated with the events of random experiments:

- permutation
- combination
- distinguishable permutation

Assumption: a random experiment can be done by a sequential implementation of two or more sub-experiments.

Multiplication Principle

Problem

Consider that an experiment E can be done by a sequential implementation of 2 sub-experiments E_1 and E_2 .

$$ightarrow$$
 Experiment E_1 $ightarrow$ n_1 outcomes

$$ightarrow$$
 Experiment E_2 $ightharpoonup n_2$ outcomes

$$ightarrow$$
 Experiment E_1 $ightharpoonup$ Experiment E_2 $ightharpoonup$ $ighthar$

Example 1

E: Test drugs A, B and placebo on rats.

 E_1 : select a rat from the cage which is either male or female,

 $n_1 = 2$

 E_2 : for each selected rat either drug A, drug B or placebo, $n_2=3$

In total there are $n_1 \cdot n_2 = 2 \times 3 = 6$ outcomes.

Then the outcomes for the experiment are denoted by

ordered pair:
$$(F,A)$$
, (F,B) , (F,P) in total $6 = 2 \times 3$

Permutation of *n* objects

Problem

Consider that n positions are to be filled with n different objects.

The task can be handled by multiplication principle.

in total $n! = n(n-1) \cdots 2 \cdot 1$ arrangements (0! = 1)

Definition: each of the n! arrangements of n different objects is called a **permutation of** n **objects**

Permutation of n objects taken r at a time

Problem

Consider that only r positions are to be filled with objects selected from n different objects.

By multiplication principle

in total ${}_{n}P_{r}=n(n-1)\cdots(n-r+1)=\frac{n!}{(n-r)!}$ arrangements.

Definition: Each of the ${}_{n}P_{r}$ arrangements is called a **permutation** of n objects taken r at a time.

Example 2

The number of possible 4-English letter words with different letters

$$_{26}P_4 = 26 \times 25 \times 24 \times 23 = \frac{26!}{22!}$$