Sample Space and Probability

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1 Sets (Quick Review)

Set, element, empety set \emptyset , finit set, countably finit set, uncountable set, sub set, equal, universal set Ω

1.1 Set opeartions:

- 1. Complement of a set S, with respect it the universe Ω , denoted by S^c
- 2. Union of two sets $S, T, S \cup T$
- 3. Intersection of two sets $S, T, S \cap T$
- 4. Union of several, $\bigcup_{n=1}^{\infty} S_n = S_1 \cup S_2 \cup ...$ 5. Intersection of several, $\bigcap_{n=1}^{\infty} S_n = S_1 \cap S_2 \cap ...$
- 6. Sets are Disjoint if they share no element
- 7. A collection of sets is a partition of set S, if they are disjoint and the union of them are S

1.2 The Algrebra of Sets:

De Morgan's laws:

$$\left(\bigcup_{n} S_{n}\right)^{c} = \bigcap_{n} S_{n}^{c}$$

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2 Probabilistic Models

Elements of a Probabilistic Model

• The sample space Ω , the set of all possiable outcomes

• The probability law, which assigns any event A a non-negative number P(A)

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2.1 Choosing an Appropriate Sample Space

The element of the sample space should be distinct and *mutually exclurive*, and the sample space should be collectively exhaustive.

2.2 Probability Axioms

1. (Nonnegativity) $P(A) \ge 0$, for every event A

2. **(Additivity)** A, B are disjoint, then $P(A \cup B) = P(A) + P(B)$

3. (Normalization) $P(\Omega) = 1$

2.3 Discrete Models

e.g. The toss of a coin several times Like {HHH,HHT,HTH,HTT,THH,THT,TTH,TTT}(3 times) and the probability stuff

Discrete Probability Law

The sample space $S = \{s_1, s_2, s_3, ..., s_n\}$ consists of finite number of elements, we have:

$$P(S) = P(\{s_1, s_2, s_3, ..., s_n\}) = P(s_1) + P(s_2) + P(s_3) + ... + P(s_n)$$

Discrete Uniform Probability Law

Ii the outcomes are equally likely, then the Probability of any single outcome A becomes:

$$P(A) = \frac{\text{number of elements of } A}{n}$$

2.4 Continuous Models

Like throughing a dart on a certian area or sth else ...

2.5 Properties of Probability Laws

1. If $A \in B$, then $P(A) \le P(B)$

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

3. $P(A \cup B) \le P(A) + P(B)$

4. $P(A \cup B \cup C) = P(A) + P(A^c \cap B) + P(A^c \cap B^c \cap C)$

^{&#}x27;Insight of Probability: The term "probability" should come with an event, like the probability of event A P(A), which is further a outcome of the probability law and a part of the probabilistic model. And a valid probabilistic model should contain a sample space and a probability law which agree with the probability axioms.

3 Conditional Prabability

Conditional probability provides us with a way to reason about the outcome of an experiment, based on **parcial information**. (The experiment is done and the only have some parcial information about it.)

e.g. The experiment involving two successive rolls of a die, you are toled that the sum of the two rolls are 9. What's the probability of the first roll is a 6?

In precise terms, the conditional probability is when we know the is with in a given event B, we wish to know the probability of the event A. We call this *conditional probability of* A *given* B, denoted by $P(A \mid B)$

Definition_{conditional probability}:
$$P(A \mid B) = \frac{P(A \cap B)}{P(B)}$$

3.1 Verification of the Prabability Laws

- 1. Nonnegativity is clear since the original probability is nonnegative.
- 2. Additivity:

$$\begin{split} P(A_1 \cup A_2 \mid B) &= \frac{P((A_1 \cup A_2) \cap B)}{P(B)} \\ &= \frac{P(A_1 \cap B) + P(A_2 \cap B)}{P(B)} \\ &= P(A_1 \mid B) + P(A_2 \mid B) \end{split}$$

1. Normalization:

$$P(\Omega \mid B) = \frac{P(\Omega \cap B)}{P(B)} = \frac{P(B)}{P(B)} = 1$$

3.2 Properties of Conditional Probability

- Conditional probability can be viewed as a normal probability on a new universe *B*.
- Furthermore, if all outcomes are equally likely, then $P(A \mid B) = \frac{\text{num of elements of } A \cup B}{\text{num of elements of } B}$

3.3 Using Conditional Probability for Modeling

A restatement of the definition of the conditional probability is $P(A \cup B) = P(B)P(A \mid B)$, which can be used to calculate a non-conditional probability.

Multiplication Rule

By definition, it's easy to get

$$P\left(\bigcup_{i=1}^{n}A_{i}\right) = P(A_{1})P(A_{2}\mid A_{3})P(A_{3}\mid A_{1}\cap A_{2})...P\left(A_{n}\mid \bigcup_{i=1}^{n-1}A_{i}\right)$$

3.4 Total Probability Theorem and Bayes' Rule