STA2001 Probability and Statistics (I)

Lecture 4

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Review

Conditional probability of an event A, given that event B has occurred, is defined by

$$P(A|B) = \frac{P(A \cap B)}{P(B)}$$

provided that P(B) > 0. Note: conditional probability is a probability function.

Events A and B are independent if

$$P(A \cap B) = P(A)P(B).$$

The occurrence of one of them does not change the probability of the occurrence of the other.

Properties of Independent Events

Theorem 1.4-1

A and B are independent, if and only if any pair of the following events are independent

- (a) A and B'
- (b) A' and B
- (c) A' and B'

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Proof:

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

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

Independent Events

Definition

Events A, B and C are mutually independent if

1. A, B, C are pairwise independent, i.e.,

$$\begin{cases} P(A \cap B) &= P(A)P(B) \\ P(A \cap C) &= P(A)P(C) \\ P(B \cap C) &= P(B)P(C) \end{cases}$$

- 2. $P(A \cap B \cap C) = P(A)P(B)P(C)$
 - multiplication rule for three independent events.

Example 3, page 39

An urn contains four balls number 1,2,3,4 and we draw one ball randomly from the urn.

$$A = \{1, 2\}, \quad B = \{1, 3\}, \quad C = \{1, 4\}$$

Then are A, B, C mutually independent?

Example 3, page 39

$$P(A) = P(B) = P(C) = \frac{1}{2}$$

$$P(A \cap B) = P(\{1\}) = \frac{1}{4} = P(A)P(B)$$

$$P(A \cap C) = P(A)P(C) = \frac{1}{4}$$

$$P(B \cap C) = P(B)P(C) = \frac{1}{4}$$

$$P(A \cap B \cap C) = \frac{1}{4} \neq P(A)P(B)P(C) = \frac{1}{8}$$

So A, B, C are pairwise independent but not mutually independent.

Mutual independence can be extended to four or more events: Each pair, triple, quartet of the events are independent and moreover

$$P(A_1 \cap \cdots \cap A_n) = P(A_1) \cdots P(A_n)$$

- ▶ If *A*, *B*, *C* are mutually independent, then
 - 1. A and $(B \cap C)$ independent,
 - 2. A' and $(B \cap C')$ independent,
 - 3. A and $(B \cup C)$ independent,
 - 4. A', B', C' independent

$$\textcircled{1}A$$
 and $(B\cap C)$ independent
$$P(A\cap (B\cap C))=P(A)P(B)P(C)=P(A)P(B\cap C)$$

(2)A' and $(B \cap C')$ independent,

By Theorem 1.4-1, $\textcircled{2} \Leftrightarrow A$ and $B \cap C'$ independent

$$P(A \cap B \cap C') = P(A \cap B) - P(A \cap B \cap C) = P(A \cap B)P(C')$$

= $P(A)P(B)P(C') = P(A)P(B \cap C')$

 $\Im A$ and $(B \cup C)$ independent

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

(4)A', B', C' independent

The pairwise independence is obvious and then from $\textcircled{3}\Leftrightarrow A'$ and $B'\cap C'$ independent

$$P(A' \cap (B' \cap C')) = P(A')P(B' \cap C') = P(A')P(B')P(C')$$

Many experiments consist of a sequence of *n* trials. If the outcomes of ith trial, in fact, does not have anything to do with the others, then events such that each is associated with a different trial should be independent in the probability sense. That is, if the event A_i is associated with the *i*th trial, $i=1,2,\cdots,n$, then A_1,A_2,\cdots,A_n are mutually independent and in particular

$$P(A_1 \cap A_2 \cap \cdots \cap A_n) = P(A_1) \cdots P(A_n)$$

Example 4, page 40

Question

A fair 6-sided die is rolled six independent times. Let $A_i = \{a \text{ match on the ith roll, i.e., the side } i \text{ is observed on the } i\text{th roll}\}, \quad i = 1, 2, \cdots, 6.$ Let $B = \{a \text{ tleast one match occur}\}$, what is P(B)?

Example 4, page 40

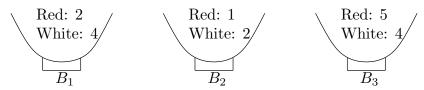
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$$P(B)=1-P(B')$$
 where $B'=\{$ no matches occur in 6 rolls $\}$
$$=1-P(A'_1\cap A'_2\cdots\cap A'_6) \quad \text{since } A'_1\cdots A'_6 \text{ are independent}$$

$$=1-P(A'_1)P(A'_2)\cdots P(A'_6)=1-\left(\frac{5}{6}\right)^6$$

Section 1.5 Bayes's Theorem



Experiment: Select a bowl first, and then draw a chip from the selected bowl.

Assumption: All chips are "equally likely" and moreover,

$$P(B_1) = \frac{1}{3}, \quad P(B_2) = \frac{1}{6}, \quad P(B_3) = \frac{1}{2}.$$

 $P(B_i)$: the probability to select the ith bowl.

Question 1

Let $R = \{ draw \ a \ red \ chip \}$. What is P(R)?

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 $=\frac{1}{3}\cdot\frac{2}{6}+\frac{1}{6}\cdot\frac{1}{3}+\frac{1}{2}\cdot\frac{5}{9}=\frac{4}{9}$

$$P(R) = P(S \cap R)$$
, where $S = \{\text{all chips}\}\$

$$= P((B_1 \cup B_2 \cup B_3) \cap R) = P((B_1 \cap R) \cup (B_2 \cap R) \cup (B_3 \cap R))$$

$$= P(B_1 \cap R) + P(B_2 \cap R) + P(B_3 \cap R)$$

$$= P(B_1)P(R|B_1) + P(B_2)P(R|B_2) + P(B_3)P(R|B_3)$$

Question 2

Suppose now that the outcome of the experiment is a red chip but we don't know from which bowl the chip was drawn. We are interested in

$$P(B_1|R), \quad P(B_2|R), \quad P(B_3|R)$$

Question 2

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From the definition of conditional probability, e.g., Consider

$$P(B_i|R) = \frac{P(B_i \cap R)}{P(R)} = \frac{P(B_i)P(R|B_i)}{P(R)}, \quad i = 1, 2, 3.$$

$$P(B_1|R) = \frac{1}{4}, \quad P(B_2|R) = \frac{1}{8}, \quad P(B_3|R) = \frac{5}{8}$$

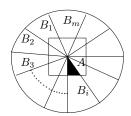
Bayes' Theorem

Assume that

- 1. S is a sample space, and B_1, B_2, \dots, B_m are mutually exclusive and exhaustive w.r.t the sample space S.
- 2. the prior probabilities of B_i is positive, i.e.,

$$P(B_i) > 0, i = 1, \cdots, m$$
. Then we have

Bayes' Theorem



(a) For any event A,

$$P(A) = \sum_{i=1}^{m} P(A \cap B_i) = \sum_{i=1}^{m} P(B_i) P(A|B_i)$$

 $\rightarrow \ total \ probability$

(b) If P(A) > 0, then

$$\begin{split} P(B_k|A) &= \frac{P(B_k \cap A)}{P(A)}, \quad k = 1, \cdots, m \\ P(B_k|A) &= \frac{P(B_k)P(A|B_k)}{P(A) = \sum_{i=1}^m P(B_i)P(A|B_i)} \rightarrow \text{Bayes Theorem} \end{split}$$

Bayes' Theorem

$$P(B_k) \rightarrow \text{ prior probability}$$

$$P(B_k|A) \rightarrow$$
 posterior probability

$$P(A|B_k) \rightarrow$$
 likelihood of B_k , A is called a data

Thomas Bayes

Thomas Bayes is known for formulating a specific case of the theorem that bears his name: Bayes' theorem.



Figure: Thomas Bayes (1701 - 1761) was an English statistician, philosopher and Presbyterian minister.

Pierre-Simon Laplace

However, it was Pierre-Simon Laplace (1749–1827) who introduced what is now called Bayes' theorem, and the Bayesian was in fact pioneered and popularised by Pierre-Simon Laplace.



Figure: Pierre-Simon Laplace (1749–1827) was a French scholar and polymath whose work was important to the development of engineering, mathematics, statistics, physics, astronomy, and philosophy.