

# STA2001 Probability and Statistics (I)

## Lecture 3

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# Review

- ▶ Revisit the method of enumeration by multiplication principle
  - ▶ permutation
  - ▶ combination
  - ▶ distinguishable permutation

## Section 1.4 Independent Events

# Motivation

## Motivation

For certain pair of events, the occurrence of one of them does not change the probability of the occurrence of the other.

## Example 1

Experiment: flip a coin twice and observe the sequence of heads and tails.

Sample space:  $S = \{HH, HT, TH, TT\}$

Assumption: the four outcomes are “equally likely”

Events:

$$A = \{\text{heads on the first flip}\} = \{HH, HT\}$$

$$B = \{\text{tails on the second flip}\} = \{HT, TT\}$$

$$C = \{\text{tails on both flips}\} = \{TT\}$$

## Example 1

$$P(A) = \frac{2}{4}, \quad P(B) = \frac{2}{4}, \quad P(C) = \frac{1}{4}$$

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Given that  $C$  has occurred, then

$$P(B|C) = 1 \text{ because } C \subset B \text{ or } \frac{P(B \cap C)}{P(C)} = \frac{P(C)}{P(C)} = 1$$

Given that  $A$  has occurred, then

$$P(B|A) = \frac{P(B \cap A)}{P(A)} = \frac{1/4}{2/4} = \frac{1}{2} = P(B)$$

Given that  $B$  has occurred, then

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

## Example 1

So we have

$$P(B|A) = P(B), \text{ and } P(A|B) = P(A)$$

the occurrence of one of them does not affect the probability of the occurrence of the other. Leading to the definition of independent events.



# Independent Events

## Definition

Events  $A$  and  $B$  are independent if

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

Otherwise, events  $A$  and  $B$  are called dependent events

- ▶ When  $P(A) \neq 0$  and  $P(B) \neq 0$ , we have

$$P(A|B) = P(A), \quad P(B|A) = P(B)$$

## Example 2, page 38

### Question

A red die and a white die are rolled.

$$S = \{(1, 1), (1, 2), \dots\}, \quad N(S) = 36$$

$$A = \{4 \text{ on the red die}\}, \quad B = \{\text{sum of dice is odd}\}$$

Assuming the two dice are fair. Are  $A$  and  $B$  independent?

## Example 2, page 38

A red die and a white die are rolled.

$$S = \{(1, 1), (1, 2), \dots\}, \quad N(S) = 36$$

$$A = \{4 \text{ on the red die}\}, \quad B = \{\text{sum of dice is odd}\}$$

Assuming the two dice are fair. Are  $A$  and  $B$  independent?

$$P(A) = \frac{6}{36}, \quad P(B) = \frac{18}{36}, \quad P(A \cap B) = \frac{3}{36}$$

$$P(A \cap B) = \frac{3}{36} = P(A)P(B) = \frac{6}{36} \cdot \frac{18}{36}$$

$\Rightarrow A$  and  $B$  are independent.

# 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:

$$\begin{aligned}P(A) &= P(A \cap (B \cup B')) = P((A \cap B) \cup (A \cap B')) \\&= P(A \cap B) + P(A \cap B') = P(A)P(B) + P(A \cap B') \\P(A \cap B') &= P(A)(1 - P(B)) = P(A)P(B')\end{aligned}$$

# 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.



# Properties of Independent Events (Continued)

- ▶ 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

# Properties of Independent Events (Continued)

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, ②  $\Leftrightarrow A$  and  $B \cap C'$  independent

$$\begin{aligned} P(A \cap B \cap C') &= P(A \cap B) - P(A \cap B \cap C) = P(A \cap B)P(C') \\ &= P(A)P(B)P(C') = P(A)P(B \cap C') \end{aligned}$$

## Properties of Independent Events (Continued)

③  $A$  and  $(B \cup C)$  independent

$$\begin{aligned} P(A \cap (B \cup C)) &= P((A \cap B) \cup (A \cap C)) = P(A \cap B) + P(A \cap C) - \\ &P((A \cap B) \cap (A \cap C)) = P(A)P(B) + P(A)P(C) - P(A)P(B)P(C) = \\ &P(A)(P(B) + P(C) - P(B)P(C)) = P(A)P(B \cup C) \end{aligned}$$

④  $A', B', C'$  independent

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

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

## Properties of Independent Events (Continued)

- ▶ Many experiments consist of a sequence of  $n$  trials. If the outcomes of  $i$ th 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,  $i = 1, 2, \dots, n$ , then  $A_1, A_2, \dots, A_n$  are mutually independent and in particular

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

## Example 4, page 40

### Question

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

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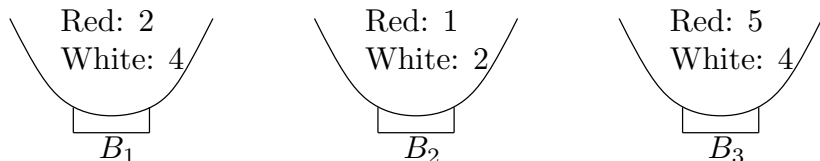
$$P(B) = 1 - P(B') \quad \text{where } B' = \{\text{no matches occur in 6 rolls}\}$$

$$= 1 - P(A'_1 \cap A'_2 \cdots \cap A'_6) \quad \text{since } A'_1 \cdots A'_6 \text{ are independent}$$

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

## Section 1.5 Bayes's Theorem

## A Motivation Example



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  $i$ th bowl.



# A Motivation Example

## Question 1

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

# A Motivation Example

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

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

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

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

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

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

# A Motivation Example

## 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

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

# A Motivation Example

## 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

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

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

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

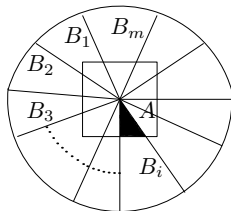
# Bayes' Theorem

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.,

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

# Bayes' Theorem



(a) For any event  $A$ ,

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

→ total probability

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

$$P(B_k|A) = \frac{P(B_k \cap A)}{P(A)}, \quad k = 1, \dots, m$$

$$P(B_k|A) = \frac{P(B_k)P(A|B_k)}{P(A) = \sum_{i=1}^m P(B_i)P(A|B_i)} \rightarrow \text{Bayes Theorem}$$

# Bayes' Theorem

$P(B_k) \rightarrow$  prior probability

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

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





# 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.