

# PROBABILITY THEORY

5

FINITE SET  $S$  OF ELEMENTARY EVENTS

$$\{ (1, b), (2, b), (3, w), (4, w) \}$$

PROBABILITY DISTRIBUTION

$$P: S \rightarrow [0, 1]$$

$$- P(s_i) \geq 0$$

$$- \sum_i P(s_i) = 1$$

- EVENT  $A$  IS ANY SUBSET OF  $S$

$$- P(A) = \sum_{s_i \in A} P(s_i)$$

$s_i \in A$

SUM OVER ELEMENTARY  
EVENTS IN  $A$

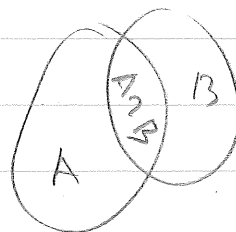
- AXIOMS:

$$\bullet P(S) = 1$$

$$\bullet P(A \cup B) = P(A) + P(B)$$

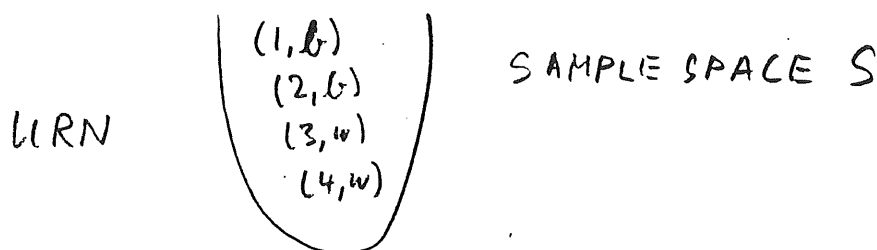
$\uparrow$   
DISJOINT  
UNION

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



$$P(A|B) = \frac{P(A \cap B)}{P(B)} \quad \text{for } P(B) > 0$$

CONDITIONAL PROB. OF  
EVENT A GIVEN EVENT B



A ball is selected from an urn containing two black balls, numbered 1 and 2, and two white balls, numbered 3 and 4. The number and color of the ball is noted, so the sample space is  $\{(1, b), (2, b), (3, w), (4, w)\}$ . Assuming that the four outcomes are equally likely, find  $P[A|B]$  and  $P[A|C]$ , where  $A$ ,  $B$ , and  $C$  are the following events:

$A = \{(1, b), (2, b)\}$ , "black ball selected,"

$B = \{(2, b), (4, w)\}$ , "even-numbered ball selected," and

$C = \{(3, w), (4, w)\}$ , "number of ball is greater than 2."

$$P(A \cap B) = P(\{(2, b)\}) = .25$$

$$P(A \cap C) = P(\emptyset) = 0$$

$$P(A|B) = \frac{P(A \cap B)}{P(B)} = \frac{.25}{.5} = .5 = P(A)$$

$$P(A|C) = \frac{P(A \cap C)}{P(C)} = \frac{0}{.5} = 0 \neq P(A)$$

In the first case, knowledge of  $B$  did not alter the probability of  $A$ . In the second case, knowledge of  $C$  implied that  $A$  had not occurred. ■■

If we multiply both sides of the definition of  $P[A | B]$  by  $P[B]$  we obtain

$$P[A \cap B] = P[A | B]P[B]. \quad (2.25a)$$

Similarly we also have that

$$P[A \cap B] = P[B | A]P[A]. \quad (2.25b)$$

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### INDEPENDENCE OF EVENTS

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If knowledge of the occurrence of an event  $B$  does not alter the probability of some other event  $A$ , then it would be natural to say that event  $A$  is independent of  $B$ . In terms of probabilities this situation occurs when

$$P[A] = P[A | B] = \frac{P[A \cap B]}{P[B]}.$$

The above equation has the problem that the right-hand side is not defined when  $P[B] = 0$ .

We will define two events  $A$  and  $B$  to be **independent** if

$$P[A \cap B] = P[A]P[B]. \quad (2.28)$$

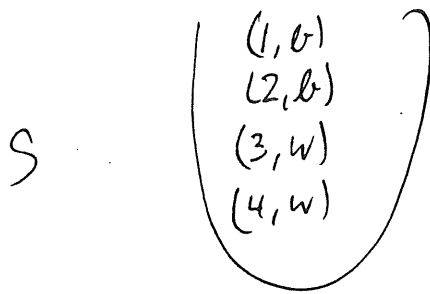
Equation (2.28) then implies both

$$P[A | B] = P[A] \quad (2.29a)$$

and

$$P[B | A] = P[B] \quad (2.29b)$$

Note also that Eq. (2.29a) implies Eq. (2.28) when  $P[B] \neq 0$  and Eq. (2.29b) implies Eq. (2.28) when  $P[A] \neq 0$ .



$A = \{(1, b), (2, b)\}$ , "black ball selected";  
 $B = \{(2, b), (4, w)\}$ , "even-numbered ball selected"; and  
 $C = \{(3, w), (4, w)\}$ , "number of ball is greater than 2."

Are events  $A$  and  $B$  independent? Are events  $A$  and  $C$  independent?

First, consider events  $A$  and  $B$ . The probabilities required by Eq. (2.28)

$$P[A] = P[B] = \frac{1}{2}, \quad \text{and}$$

$$P[A \cap B] = P[\{(2, b)\}] = \frac{1}{4}.$$

Thus

$$P[A \cap B] = \frac{1}{4} = P[A]P[B],$$

and the events  $A$  and  $B$  are independent. Equation (2.29b) gives more insight into the meaning of independence:

$$P[A | B] = \frac{P[A \cap B]}{P[B]} = \frac{P[\{(2, b)\}]}{P[\{(2, b), (4, w)\}]} = \frac{1/4}{1/2} = \frac{1}{2}$$

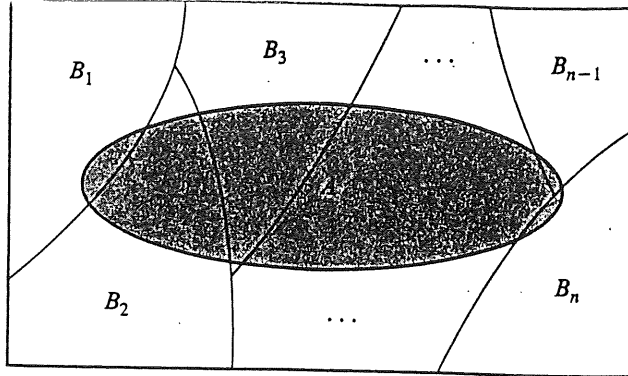
$$P[A] = \frac{P[A]}{P[S]} = \frac{P[\{(1, b), (2, b)\}]}{P[\{(1, b), (2, b), (3, w), (4, w)\}]} = \frac{1/2}{1}.$$

These two equations imply that  $P[A] = P[A | B]$  because the proportion of outcomes in  $S$  that lead to the occurrence of  $A$  is equal to the proportion of outcomes in  $B$  that lead to  $A$ . Thus knowledge of the occurrence of  $B$  does not alter the probability of the occurrence of  $A$ .

Events  $A$  and  $C$  are not independent since  $P[A \cap C] = P[\emptyset] = 0$  so

$$P[A | C] = 0 \neq P[A] = .5.$$

In fact,  $A$  and  $C$  are mutually exclusive since  $A \cap C = \emptyset$ , so the occurrence of  $C$  implies that  $A$  has definitely not occurred. ■■



**FIGURE 2.14** A partition of  $S$  into  $n$  disjoint sets.

Let  $B_1, B_2, \dots, B_n$  be mutually exclusive events whose union equals the sample space  $S$  as shown in Fig. 2.14. We refer to these sets as a **partition** of  $S$ . Any event  $A$  can be represented as the union of mutually exclusive events in the following way:

$$\begin{aligned} A &= A \cap S = A \cap (B_1 \cup B_2 \cup \dots \cup B_n) \\ &= (A \cap B_1) \cup (A \cap B_2) \cup \dots \cup (A \cap B_n). \end{aligned}$$

See Fig. 2.14. By Corollary 4, the probability of  $A$  is

$$P[A] = P[A \cap B_1] + P[A \cap B_2] + \dots + P[A \cap B_n].$$

By applying Eq. (2.25a) to each of the terms on the right-hand side, we obtain the **theorem on total probability**:

$$P[A] = P[A | B_1]P[B_1] + P[A | B_2]P[B_2] + \dots + P[A | B_n]P[B_n].$$

KNOWLEDGE OF  $P(A | B_i)$

AND  $P(B_i)$

LETS US COMPUTE  $P(A)$

## Bayes' Rule

Let  $B_1, B_2, \dots, B_n$  be a partition of a sample space  $S$ . Suppose that event  $A$  occurs, what is the probability of event  $B_j$ ? By the definition of conditional probability we have

$$P[B_j | A] = \frac{P[A \cap B_j]}{P[A]} = \frac{P[A | B_j]P[B_j]}{\sum_{k=1}^n P[A | B_k]P[B_k]}, \quad (2.27)$$

where we used the theorem on total probability to replace  $P[A]$ . Equation (2.27) is called **Bayes' rule**.

$P(B_j)$  PRIOR PROBABILITIES

EXPERIMENT PERFORMED AND  
A OCCURRED

$P(B_j | A)$  POSTERIOR PROBABILITIES  
GIVEN ADDITIONAL INFORMATION

# BAYES

- $N$  EXPERTS / MODELS  $E_i$
- IN EACH TRIAL  $t$  WE OBSERVE LABEL  $y_t$   
DATUM

## ASSUMPTION :

- ONE EXPERT  $E_i$  GENERATED  $(y_1, y_2, \dots, y_T) = \bar{y}$
- PRIOR PROBABILITY OF EXPERT  $E_i$  IS  $P(E_i)$

$y_t \in Y$  FINITE

PROBABILITY OF DATA  $\bar{y}$  GIVEN  $E_i$  GENERATED IT :

$P(\bar{y} | E_i)$

DATA LIKELIHOODS

## IMPORTANT SPECIAL CASE :

$y_1, y_2, \dots, y_T$  ARE GENERATED  
INDEPENDENTLY AT RANDOM

$$\text{THUS } P(y_1, \dots, y_T | E_i) = \prod_{t=1}^T P(y_t | E_i)$$

## GENERAL CASE

$$P(y_1, \dots, y_T | E_i) = \prod_{t=1}^T P(y_t | E_i, y_1, \dots, y_{t-1})$$

FOR EXAMPLE : EXPERTS ARE COINS  $Y = \{0, 1\}$

	$E_1$	$E_2$	$E_3$	$E_4$
$P(1 E_i)$	.1	.2	.8	.9
$P(E_i)$	.2	.4	.3	.1

$$\bar{y}_3 = (1, 1, 0)$$

$$P(E_i | \bar{y}_3) = \frac{P(\bar{y}_3 | E_i) P(E_i)}{P(\bar{y}_3)}$$

POSTERIOR

$$= \frac{P(1|E_i)^2 (1 - P(1|E_i)) P(E_i)}{P(\bar{y}_3)}$$

	$E_1$	$E_2$	$E_3$	$E_4$
$P(E_i   \bar{y}_3) \sim$	$.1^2 \cdot .9 \cdot .2$	$.2^2 \cdot .8 \cdot .4$	$.8^2 \cdot .2 \cdot .3$	$.9^2 \cdot .1 \cdot .1$
$\sim$	18	128	384	81

FOR 1-HEAVY SEQUENCES

POSTERIOR WILL BECOME  $\approx \underset{i}{\operatorname{argmax}} P(1|E_i)$

PROVIDED THAT ALL  $P(E_i) > 0$