# 0.1 Additive Rules

• Theorem 2.10:

If A and B are any two events, then

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

# • Corollary 1:

If A and B are mutually exclusive, then

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



Figure 1: Additive rule of probability.

• Corollary 2:

If 
$$
A_1, A_2, \ldots A_n
$$
, are mutually exclusive, then  
\n
$$
P(A_1 \cup A_2 \cup \ldots \cup A_n) = P(A_1) + P(A_2) + \ldots + P(A_n)
$$

• Corollary 3:

If 
$$
A_1, A_2, \ldots A_n
$$
, is a partition of a sample space *S*, then  
\n
$$
P(A_1 \cup A_2 \cup \ldots \cup A_n) = P(A_1) + P(A_2) + \ldots + P(A_n)
$$
\n
$$
= P(S) = 1
$$

• Theorem 2.11: (an extension of Theorem 2.10)

For three events  $A, B$ , and  $C$ ,  $P(A \cup B \cup C) = P(A) + P(B) + P(C) - P(A \cap B) - P(A \cap C)$  $-P(B\cap C) + P(A\cap B\cap C)$ 

### • Theorem 2.12:

If  $\overline{A}$  and  $\overline{A}'$  are complementary events, then

$$
P(A) + P(A') = 1
$$

Proof : Since  $A \cup A' = S$  and  $A \cap A' = \emptyset$ , then

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

- **Example 2.32**: The probability that the production procedure meets specification  $(2000 \pm 10 \; mm)$  is known to be 0.99. Small cable is just as likely to be defective as large cable.
	- What is the probability that a cable selected randomly is too large?

Let M be the event that a cable meets spec. Let  $S$  and  $L$  be the events that the cable is too small and too large, respectively. Then

 $P(M) = 0.99$  and  $P(S) = P(L) = (1 - 0.99)/2 = 0.0005$ 

– What is the probability that a cable selected randomly is larger than 1990 mm?

 $P(X > 1990) = 1 - P(S) = 0.995$ 

where  $X$  is the length of a randomly selected cable.

### 0.2 Conditional Probability

- Conditional probability:  $P(B|A)$ 
	- Sometimes the occurrence of an event is influenced or related with some other event.
	- Hence we must take this restriction or the availability of certain limited information into consideration about the outcome of the experiment.
- $-$  The probability of an event B occurring when it is known that some event A has occurred.
- "The probability that  $B$  occurs given that  $A$  occurs" or "The probability of  $B$ , given  $A$ "
- The notion of conditional probability provides the capability of reevaluating the idea of probability of an event in light of additional information.
- Example:  $S = 1, 2, 3, 4, 5, 6, A = 4, 5, 6, B = 1, 3, 5, \Rightarrow P(B|A)$ ?
- Definition 2.9:

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

provided  $P(A) > 0$ 

• Example: Our sample space  $S$  is the population of adults in a small town. They can be categorized according to gender and employment status (see Table [1\)](#page-2-0).

<span id="page-2-0"></span>Table 1: Categorized adult population in a small town.



- One individual is to be selected at random for a publicity tour.
- The concerned events
	- $M: a man is chosen$
	- $E$ : the one chosen is employed

$$
P(M|E) = \frac{460}{600} = \frac{23}{30}
$$

$$
P(M|E) = \frac{n(E \cap M)/n(S)}{n(E)/n(S)} = \frac{P(E \cap M)}{P(E)} = \frac{\frac{460}{900}}{\frac{600}{900}} = \frac{23}{30}
$$

• **Example 2.33**: The probability that a regularly scheduled flight departs on time is  $P(D) = 0.83$ ;

- the probability that arrives on time is  $P(A) = 0.82$ ;
- the probability that it departs and arrives on time is  $P(D \cap A) = 0.78$ .
- Find the probability that a plane
	- arrives on time given that it departed on time, and

$$
P(A|D) = \frac{P(D \cap A)}{P(D)} = \frac{0.78}{0.83} = 0.94
$$

– departed on time given that it has arrived on time.

$$
P(D|A) = \frac{P(D \cap A)}{P(A)} = \frac{0.78}{0.82} = 0.95
$$



- Definition 2.
	- $P(B|A) = P(B)$  or  $P(A|B) = P(A)$ .

Otherwise,  $A$  and  $B$  are **dependent**.

- If knowing that event  $B$  occurred doesn't change the probability that A will occur, then B must carry no information about A.
- The condition  $P(B|A) = P(B)$  implies that  $P(A|B) = P(A)$ , and conversely.
- Example: Two cards are drawn in succession, with replacement
	- Event A: the first card is an ace
	- $-$  Event  $B$ : the second card is a spade

$$
P(B|A) = \frac{P(A \cap B)}{P(A)} = \frac{1/52}{4/52} = \frac{13}{52} = \frac{1}{4} \text{ and } P(B) = \frac{13}{52} = \frac{1}{4}
$$

– Since  $P(B|A) = P(B)$ , these two events are independent.

# 0.3 Multiplicative Rules

• Multiplying the formula of Definition 2.9 by  $P(A)$ , we obtain the **mul**tiplicative rule, which enables us to calculate the probability that two events will both occur.

Theorem 2.13:

If in an experiment the events  $A$  and  $B$  can both occur, then

$$
P(A \cap B) = P(A) * P(B|A)
$$

provided  $P(A) > 0$ 

• We can also write

$$
P(A \cap B) = P(B \cap A) = P(B) * P(A|B)
$$

- **Example 2.35**: Suppose that we have a fuse box containing 20 fuses, of which 5 are defective. If 2 fuses are selected at random and removed from the box in succession without replacing the first.
- What is the probability that both fuses are defective?
	- Event A: the first fuse is defective
	- $-$  Event  $B$ : the second fuse is defective. Hence,

$$
P(A \cap B) = P(A) * P(B|A) = \frac{1}{4} * \frac{4}{19} = \frac{1}{19}
$$

- Example 2.36: One bag contains 4 white balls and 3 black balls. A second bag contains 3 white balls and 5 black balls.
- One ball is drawn from the first bag and placed unseen in the second bag. What is the probability that a ball now drawn from the second bag is black?
- Solution: Let  $B_1, B_2$ , and  $W_1$  represent, respectively, the drawing of a black ball from bag 1, a black ball from bag 2, and a white ball from bag 1.

$$
p[(B_1 \cap B_2) \cup (W_1 \cap B_2)] = P(B_1 \cap B_2) + P(W_1 \cap B_2)
$$
  
=  $P(B_1)P(B_2|B_1) + P(W_1)P(B_2|W_1)$   
=  $\frac{3}{7} * \frac{6}{9} + \frac{4}{7} * \frac{5}{9} = \frac{38}{63}$ 

#### Theorem 2.14:

Two events  $A$  and  $B$  are (statistically or probabilistically) independent if and only if

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

. Therefore, to obtain the probability that two independent events will both occur, we simply find the product of their individual probabilities.



Figure 2: Tree diagram for Example 2.36.

- Example 2.37: A small town has one fire engine and one ambulance available for emergencies.
	- The probability that the fire engine is available when needed is 0.98,
	- The probability that the ambulance is available when called is 0.92
	- In the event of an injury resulting from a burning building, find the probability that both the ambulance and the fire engine will be available.
- Solution: Let  $A$  and  $B$  represent the respective evens that the fire engine and the ambulance are available. Then

$$
P(A \cap B) = P(A)P(B) = 0.98 \times 0.92 = 0.9016.
$$

- Example 2.38: Find the probability that
	- the entire system works
	- the component  $C$  does not work, given that the entire system works
- Solution:

$$
P(A \cap B \cap (C \cup D)) = P(A) * P(B) * P(C \cup D)
$$



Figure 3: An electrical system for Example 2.38.

$$
= P(A) * P(B) * (1 - P(C' \cap D')) = P(A) * P(B) * (1 - P(C') * P(D'))
$$

$$
= 0.9 * 0.9 * (1 - (1 - 0.8) * (1 - 0.8)) = 0.7776
$$

$$
\theta
$$

$$
P = \frac{P(the\ system\ works\ but\ C\ does\ not\ work)}{P(the\ system\ works)}
$$

$$
= \frac{P(A \cap B \cap C' \cap D)}{P(A \cap B \cap (C \cup D))} = \frac{0.9 * 0.9 * (1 - 0.8) * 0.8}{0.7776} = 0.1667
$$

- Independence is often easy to grasp intuitively.
- For example, if the occurrence of two events is governed by distinct and non-interacting physical processes, such events will turn out to be independent.
- On the other hand, independence is not easily visualized in terms of the sample space.
- A common fallacy (wrong idea) is that two events are independent if they are disjoint, but in fact the opposite is true:

Two disjoint events A and B with  $P(A) > 0$  and  $P(B) > 0$  are never independent, since their intersection  $A \cap B$  is empty and has probability 0.

- We note that
	- (i) independent events are never mutually exclusive,
	- (ii) two mutually exclusive events are always dependent.

Theorem 2.15:

**Theorem 2.15:**<br>If the events  $A_1, A_2, A_3, \ldots, A_k$  can occur, then  $P(A_1 \cap A_2 \cap ... \cap A_k) = P(A_1)P(A_2|A_1)P(A_3|A_1 \cap A_2)$  $\ldots P(A_k|A_1 \cap A_2 \cap \ldots \cap A_k)$ If the events  $A_1, A_2, A_3, \ldots, A_k$  are independent, then

$$
P(A_k|A_1 \cap A_2 \cap \ldots \cap A_k) = P(A_1)P(A_2) \ldots P(A_k) = \prod_{n=1}^k P(A_n)
$$

- Example 2.39: Three cards are drawn in succession without replacement. Find the probability that the event  $A_1 \cap A_2 \cap A_3$  occurs, where
	- $A_1$ : the first card is red ace
	- $A_2$ : the second card is a 10 or jack
	- $A_3$ : the third card is greater than 3 but less than 7
- Solution:

$$
P(A_1 \cap A_2 \cap A_3) = P(A_1)P(A_2|A_1)P(A_3|A_1 \cap A_2)
$$
  
=  $\frac{2}{52} * \frac{8}{51} * \frac{12}{50} = \frac{8}{5525}$ 

#### • Independence of Several Events:

The events  $A_1, A_2, A_3, \ldots, A_n$  are **independent** if  $P(\bigcap$ i∈S  $A_i$ ) =  $\prod$ i∈S  $P(A_i)$ 

for any subset S of  $\{1, 2, \ldots, n\}$ .

- Independence means that the occurrence or non-occurrence of any number of the events from that collection carries no information on the remaining events or their complements.
- Example: Independence of three events: If  $A_1, A_2$  and  $A_3$  are independent,

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

$$
P(A_1 \cap A_3) = P(A_1)P(A_3)
$$

$$
P(A_2 \cap A_3) = P(A_2)P(A_3)
$$
  

$$
P(A_1 \cap A_2 \cap A_3) = P(A_1)P(A_2)P(A_3)
$$

- Example: Consider two independent fair coin tosses, and the following events:
	- $H_1$  = 1<sup>st</sup> toss is a head,
	- $H_2 = 2^{nd}$  toss is a head,
	- $-D =$  the two tosses have different results.
- Pairwise independence does not imply independence.
	- $H_1$  and  $H_2$  are independent, by definition.
	- $P(D|H_1) = P(D)$  and  $P(D|H_2) = P(D)$
	- $-P(H_1 \cap H_2 \cap D) = 0 \neq P(H_1)P(H_2)P(D)$
- Example: Consider two independent rolls of a fair die, and the following events:

–  $A = 1^{st}$  roll is 1, 2, or 3,  $B = 2^{nd}$  roll is 3, 4, or 5,  $C =$  the sum of the two rolls is 9.

- $P(A_1 \cap A_2 \cap A_3) = P(A_1)P(A_2)P(A_3)$  is not enough for independence.
	- $-P(A \cap B) = \frac{1}{6} \neq \frac{1}{2}$  $\frac{1}{2} * \frac{1}{2} = P(A)P(B)$
	- $-P(A \cap C) = \frac{1}{36} \neq \frac{1}{2}$  $\frac{1}{2} * \frac{4}{36} = P(A)P(C)$
	- $-P(B \cap C) = \frac{3}{6} \neq \frac{1}{2}$  $\frac{1}{2} * \frac{4}{36} = P(B)P(C)$
	- $-P(A \cap B \cap C) = \frac{1}{36} \neq \frac{1}{2}$  $\frac{1}{2} * \frac{1}{2}$  $\frac{1}{2} * \frac{4}{36} = P(A)P(B)P(C)$

## 0.4 Bayes'Rules

- Our sample space S is the population of adults in a small town. They can be categorized according to employment status.
- One individual is to be selected at random for a publicity tour.
	- $-$  The concerned event  $E$ : the one chosen is employed
	- Give the additional information that 36 of those employed and 12 of those unemployed are members of the Rotary Club.
	- $-$  Find the probability of the event A that individual selected is a member of the Rotary Club.



Figure 4: Venn diagram for the events  $A, E$ , and  $E'$ .

• Event A is the union of the two mutually exclusive events  $E \cap A$  and  $E' \cap A$ . Hence,

•
$$
A = (E \cap A) \cup (E' \cap A)
$$
  
\n•
$$
P(A) = P[(E \cap A) \cup (E' \cap A)]
$$
  
\n
$$
= P(E \cap A) + P(E' \cap A)
$$
  
\n
$$
= P(E)P(A|E) + P(E')P(A|E')
$$
  
\n•
$$
P(E) = \frac{600}{900} = \frac{2}{3}, \ P(A|E) = \frac{36}{600} = \frac{3}{50}
$$
  
\n•
$$
P(E') = \frac{1}{3}, P(A|E) = \frac{12}{300} = \frac{1}{25}
$$
  
\n•
$$
P(A) = \frac{2}{3} * \frac{3}{50} + \frac{1}{3} * \frac{1}{25} = \frac{4}{75}
$$

• Theorem 2.16: (Theorem of total probability or rule of elimination)

If the events  $B_1, B_2, \ldots, B_k$  constitute a partition of the sample space S such that  $P(B_i) \neq 0$  for  $i = 1, 2, ..., k$ , then for any event A of S,  $P(A) = \sum$ k  $i=1$  $P(B_i \cap A) = \sum$ k  $i=1$  $P(B_i)P(A|B_i)$ 



Figure 5: Tree diagram for the data.



Figure 6: Partitioning the sample space S.

- Example 2.41: In a certain assembly plant, three machines,  $B_1, B_2$ and  $B_3$  make  $30\%,\,45\%$  and  $25\%,$  respectively, of the products.
- It is known from past experience that  $2\%, 3\%,$  and  $2\%$  of the products made by each machine, respectively, are defective.
- Now, suppose that a finished product is randomly selected. What is the probability that it is defective?

$$
P(A) = P(B_1)P(A|B_1) + P(B_2)P(A|B_2) + P(B_3)P(A|B_3)
$$
  
= 03 \* 0.02 + 0.45 \* 0.03 + 0.25 \* 0.02 = 0.0245

- Solution:
- Event A: the product is defective.
- Event  $B$ : the product is made by machine  $B_i$



Figure 7: Tree diagram for Example 2.41.

• Theorem 2.17: (Bayes'Rule)

If the events  $B_1, B_2, \ldots B_k$  constitute a partition of the sample space S such that  $P(B_i) \neq 0$  for  $i = 1, 2, \ldots, k$ , then

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

• It can be proved by the definition of conditional probability,

$$
P(B_r|A) = P(B_r \cap A)/P(A)
$$

and then using Theorem 2.16 in the denominator.

- Useful in problems where  $P(B_i|A)$  are not known but  $P(A|B_i)$  and  $P(B_i)$  are known.
- Some terminology:
	- $P(B_i)$ : priors
	- $P(A|B_i)$ : likelihoods
	- $P(B_i|A)$ : posteriors
- Example 2.42: With reference to Example 2.41, if a product were chosen randomly and found to be defective, what is the probability that it was made by machine  $B_3$
- Using Bayes'rule,

$$
P(B_3|A) = \frac{P(B_3)P(A|B_3)}{P(B_1)P(A|B_1) + P(B_2)P(A|B_2) + P(B_3)P(A|B_3)}
$$

$$
= \frac{0.005}{0.006 + 0.0135 + 0.005} = \frac{10}{49}
$$

- Example 2.43: A manufacturing firm employs three analytical plans for the design and development of a particular product.
- For cost reasons, all three are used at varying times. In fact, plans 1, 2, and 3 are used for 30%, 20%, and 50% of the products respectively.
- The "defect rate" is different for the three procedures as follows:

$$
P(D|P_1) = 0.01
$$
,  $P(D|P_2) = 0.03$ ,  $P(D|P_3) = 0.5$ 

where  $P(D|P_i)$  is the probability of a defective product, given plan j.

- If a random product was observed and found to be defective, which plan was most likely used and thus responsible?
- Solution:  $P(P_1) = 0.3, (P_1 2) = 0.2, (P_1) = 0.5$

$$
P(P_i|D) = \frac{P(P_i)P(D|P_i)}{\sum_{i=1}^{3} P(P_i)P(D|P_i)} = \frac{(0.30)(0.01)}{(0.3)(0.01) + (0.20)(0.03) + (0.50)(0.02)} = \frac{0.003}{0.019}
$$

$$
P(P_1|D) = 0.158
$$
,  $P(P_2|D) = 0.316$ ,  $P(P_3|D) = 0.526$ .