

## 1.5 Independence

### Definition

**Independent Events** Events  $A$  and  $B$  are **independent** if  $P(A \cap B) = P(A)P(B)$ .

Equivalently,  $A$  and  $B$  are independent if

$$P(A \mid B) = P(A),$$

assuming  $P(B) > 0$ .

The first definition is preferred because it can be used even when  $P(A \mid B)$  is undefined.

## Implications

Practically, two events are independent if knowledge about one of the events occurring has no impact the probability of the other event occurring.

If  $A$  is independent of  $B$ , then  $B$  is independent of  $A$ .

Shockingly, disjoint events cannot be independent!

Why? If  $P(A) > 0$  and  $P(B) > 0$  but  $A$  and  $B$  are disjoint, then  $0 = P(A \cap B) \neq P(A)P(B) > 0$ .

### Example 1.19

Consider an experiment involving two successive rolls of a 4-sided die in which all 16 possible outcomes are equally likely and have probability  $1/16$ .

### Example 1.19 (cont)

Are the events below independent?

$A_i = \{\text{1st roll results in } i\}$ ,  $B_j = \{\text{2nd roll results in } j\}$

### Example 1.19 (cont)

Are the events below independent?

$A_i = \{\text{1st roll is a } i\}$ ,  $B_j = \{\text{sum of the two rolls is } 5\}$

### Example 1.19 (continued)

Are the events below independent?

$A_i = \{\text{maximum of the two rolls is } 2\},$

$B_j = \{\text{minimum of the two rolls is } 2\}$

### Conditional Independence

## Definition

### Conditionally Independent Events

Given an event  $C$ , the events  $A$  and  $B$  are **conditionally independent** if  $P(A \cap B \mid C) = P(A \mid C)P(B \mid C)$ .

The independence of  $A$  and  $B$  does not imply the conditional independence of the events.

## Alternative Characterization

### Example 1.20

Consider two independent fair coin tosses, in which all four possible outcomes are equally likely. Let

$\{H_1\}$  = 1st toss is a head,

$\{H_2\}$  = 2nd toss is a head,

$D = \{\text{the two tosses have different results}\}.$

### Example 1.20 (cont)

### Example 1.21

There are two coins, a blue and red one. We choose one of the two coins at random, each being chosen with probability  $1/2$ . We proceed with two independent tosses. The coins are biased: with the blue coin, the probability of heads in any given toss is 0.99, whereas for the red coin it is 0.01.

Let  $B$  be the event that the blue coin was selected. Let  $H_i$  be the event that the  $i$ th toss results in a head.

The events  $H_1$  and  $H_2$  are dependent. However, they are conditionally independent given the coin.

### Example 1.21 (cont)

## Summary

- Two event  $A$  and  $B$  are **independent** if  $P(A \cap B) = P(A)P(B)$ .
- If  $A$  and  $B$  are independent, then  $A$  and  $B^c$  are independent.
- Two events  $A$  and  $B$  are **conditionally independent** given another event  $C$  with  $P(C) > 0$ , if  $P(A \cap B \mid C) = P(A \mid C)P(B \mid C)$ .
- Independence does not imply conditional independence or vice versa.

## Independence of a Collection of Events



## Independence Generalization

### Definition of Independence of Several Events

The events  $A_1, A_2, \dots, A_n$  are independent if

$$P(\cap_{i \in S} A_i) = \prod_{i \in S} P(A_i), \quad \text{for every subset } S \subseteq \{1, 2, \dots, n\}.$$

## Independence Generalization

If  $A_1, A_2$ , and  $A_3$  are events, what properties do they have to satisfy to be independent?

### Example 1.22

Pairwise independence does not imply independence.

Consider two independent fair coin tosses, in which all four possible outcomes are equally likely. Let

$\{H_1\}$  = 1st toss is a head,

$\{H_2\}$  = 2nd toss is a head,

$D = \{\text{the two tosses have different results}\}.$

### Example 1.22 (cont)

### Example 1.23

The equality  $P(A_1 \cap A_2 \cap A_3) = P(A_1)P(A_2)P(A_3)$  does not guarantee independence.

Consider two rolls of a fair six-sided die, and the following events:

$A = \{\text{1st roll is 1, 2, or 3}\},$

$B = \{\text{1st roll is 3, 4, or 5}\},$

$C = \{\text{the sum of the two rolls is 9}\}.$

### Example 1.23 (cont)

## Intuition

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.

## Reliability

## Context

In probabilistic models of complex systems involving several components, it is often convenient (and intelligent) to assume that the behaviors of the components are uncoupled (independent).

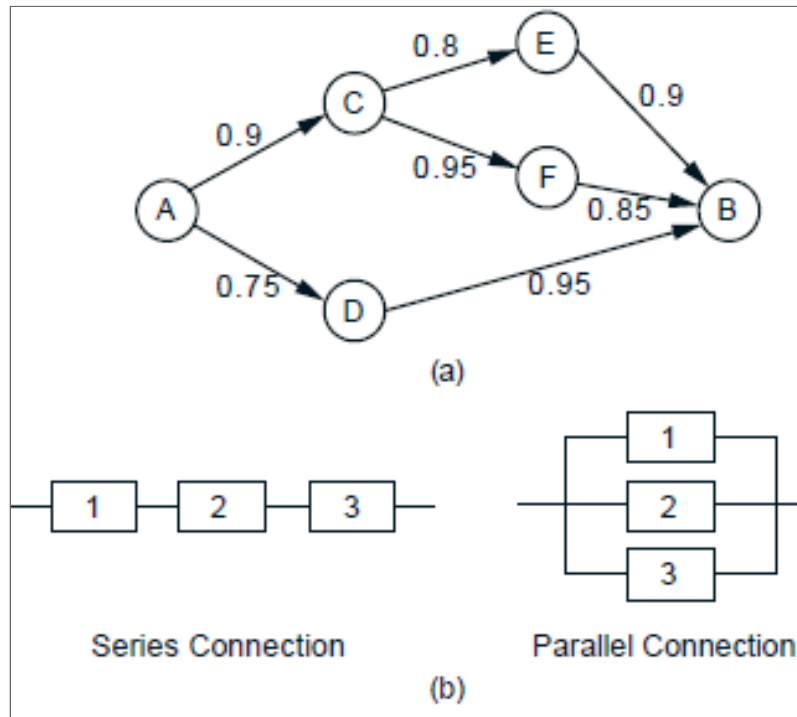
### Example 1.24 Network Connectivity

A computer network connects two nodes  $A$  and  $B$  through intermediate nodes  $C, D, E, F$ .

For every pair of directly connected nodes, say  $i$  and  $j$ , there is a given probability  $p_{ij}$  that the link from  $i$  to  $j$  is functioning properly. We assume that link failures are independent of each other.

What is the probability that there is a path connecting  $A$  and  $B$  in which all links are up?

### Example 1.24 (cont)



Reliability example

### **Example 1.24 (cont)**

## **Independent Trials and the Binomial Probabilities**

## Context

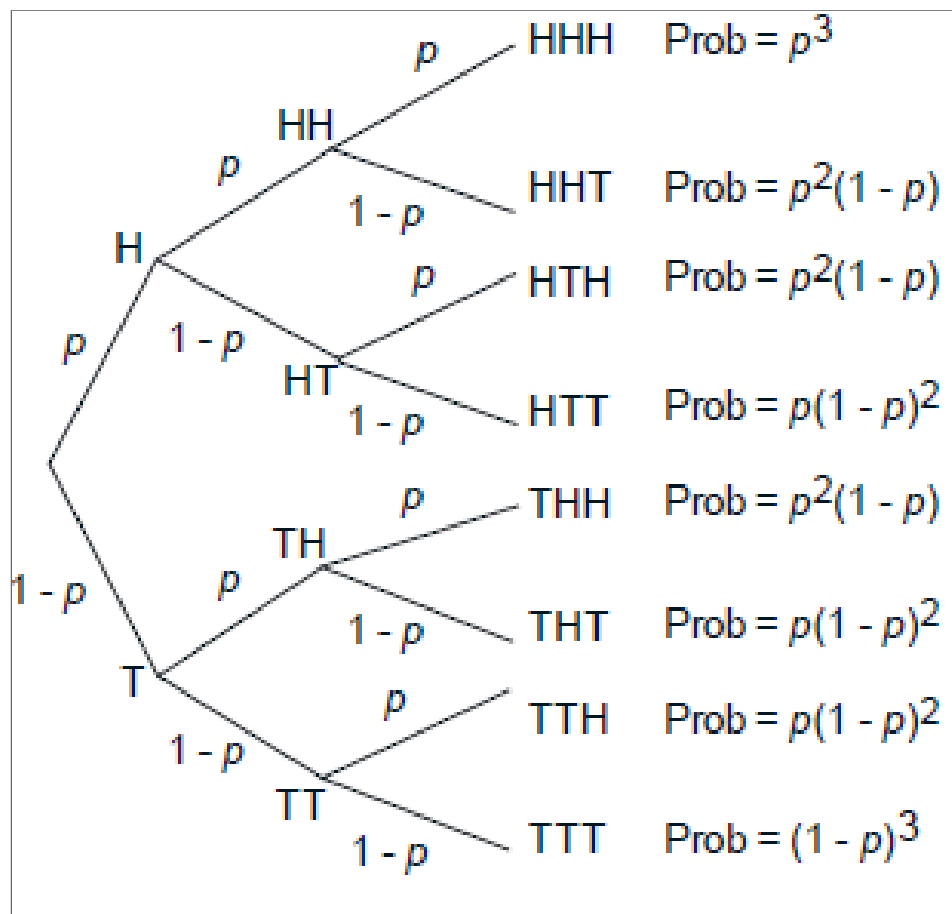
If an experiment involves a sequence of identical smaller experiments, then we have a sequence of **independent trials**.

If each trial has only two possible outcomes, then we have a sequence of **Bernoulli trials**.

Assume we flip a coin three times, successively and independently. Let  $H_i$  denote flipping a head on trial  $i$  and  $P(H_i) = p$ .

## Bernoulli trials visualized





Bernoulli trials tree diagram

## Binomial probabilities

For a set of Bernoulli trials, consider the probability

$p(k) = P(k \text{ heads come up in an } n\text{-toss sequence})$ .

The probability for a particular sequence (one branch of the tree) is

Thus,

$p(k) =$

## Binomial probabilities

The numbers  $\binom{n}{k}$  is read as “ $n$  choose  $k$ ” and is the number of distinct  $n$ -toss sequences that contain  $k$  heads.

The numbers  $\binom{n}{k}$  are known as the **binomial coefficients**.

The numbers  $p(k)$  are known as the **binomial probabilities**.

### Example 1.25 Grade of Service

An internet service provider has installed  $c$  modems to serve the needs of a population of  $n$  dial-up customers.

Each customer will need a connection with probability  $p$  independent of the other customers.

What is the probability that there are more customers needing a connection than modems?

Answer this problem generally, and then for  $n = 100$ ,  $p = 0.1$ , and  $c = 15$ .

### Example 1.25 (cont)