

Chapter 13

# General Probability

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## Most People Guess Wrong

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When researchers pose this question, the most common answers fall between 90% and 99%.

Common guesses: 90–99%

Actual answer:

This is not about being bad at math.

Human intuition struggles with conditional probability when base rates are small.

## Why Does This Happen?

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The test is accurate, but the disease is rare. When you test positive, there are two possible explanations:

1 in 1,000

### True positive

You have the disease and the test correctly detected it.

999 in 1,000

### False positive

You do not have the disease but the test incorrectly flagged you.

Conditional probability gives us the tools to figure out which case is more likely. Let's begin.

## Intended Learning Outcomes

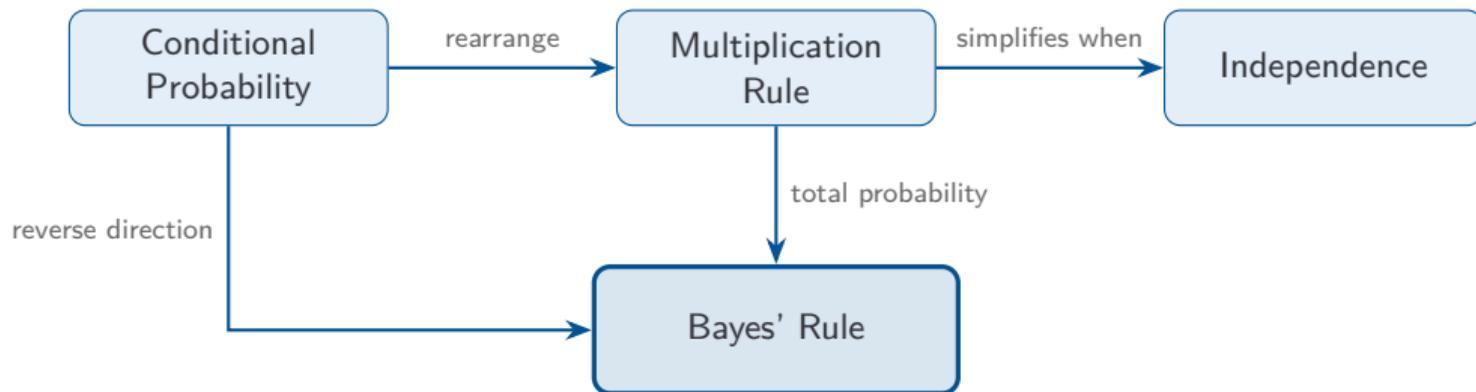
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- Define and compute conditional probabilities
- Apply the multiplication rule
- Use tree diagrams for multi-stage problems
- Define and test for independence
- Use the Law of Total Probability
- Apply Bayes' Rule to reverse conditionals
- Distinguish independence from mutual exclusivity
- Avoid common probability reasoning errors

## Chapter 13: The Big Picture

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How does learning new information change probabilities?



PART 1

# Conditional Probability

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How does knowing  $B$  occurred change the probability of  $A$ ?

## Updating Probability with New Information

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How does learning new information change what we believe?

- A patient tests positive for a disease. What is the probability they actually have it?
- A student is in first year. What is the probability they live on campus?
- You rolled an even number. What is the probability it was a 2?

 **Key Point:** Conditional probability answers: “What is the probability of  $A$ , *given that* we know  $B$  has occurred?”

# Conditional Probability: Definition

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## Conditional Probability

The **conditional probability** of event  $A$  given event  $B$  has occurred is:

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

provided that  $P(B) > 0$ .

- The bar “|” is read “given”
- Denominator  $P(B)$  restricts to outcomes where  $B$  occurred
- Conditioning “zooms in” on the subset of outcomes where  $B$  happened

## Conditional Probability: Terminology

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### Joint and Marginal Probabilities

$P(A \text{ and } B)$ : **joint probability** — the probability both events occur.

$P(A)$ ,  $P(B)$  alone: **marginal probabilities** — the probability of a single event.

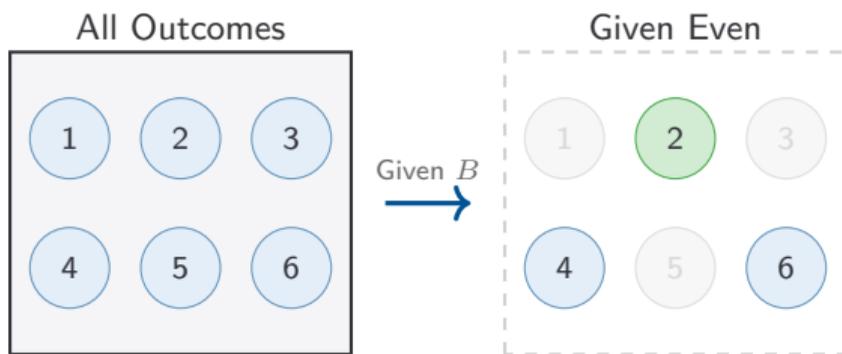
Conditioning on  $B$  means we zoom in on outcomes where  $B$  occurred and ask what fraction are also in  $A$ .



## The Key Insight: Restricting the Sample Space

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 **Key Point:** Conditional probability *restricts* our attention to only those outcomes where the condition occurred.



## Conditional Probability from a Table

### Example 13.2

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**Context:** The NCAA surveyed 3,527 young adults (ages 18–22) about sports betting.

**Calculate:** Marginal and conditional probabilities from the table.

	Bets	Does Not	Total
Male	1,182	582	1,764
Female	899	864	1,763
Total	2,081	1,446	3,527

(a)  $P(\text{Bets on sports})$

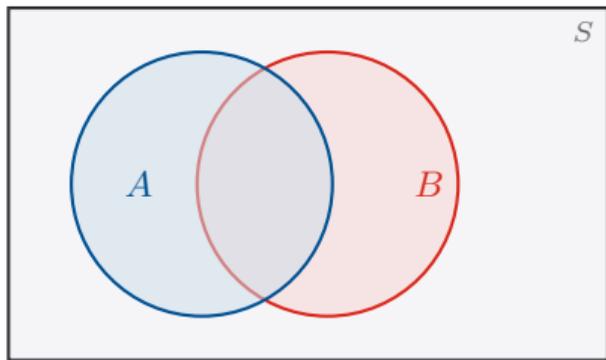
(b)  $P(\text{Bets} \mid \text{Male})$

(c)  $P(\text{Female} \mid \text{Does Not Bet})$

# Visualizing Conditional Probability

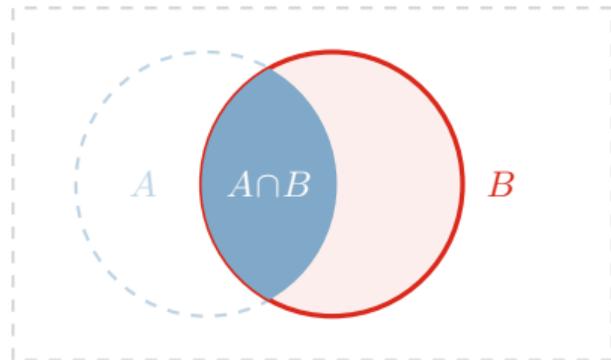
## Example 13.3

Marginal:  $P(A)$



$$P(A) = \frac{|A|}{|S|}$$

Conditional:  $P(A | B)$



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



**Key Point:** Conditioning on  $B$  replaces  $S$  with  $B$ . What fraction of  $B$  lies in  $A$ ?

PART 2

# The Multiplication Rule

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Finding joint probabilities from conditional ones

## From Conditional to Joint Probability

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If we know  $P(A | B)$ , can we find  $P(A \text{ and } B)$ ?

Starting from the definition and rearranging:

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

### Multiplication Rule

For any events  $A$  and  $B$ :

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

$P(A \text{ and } B)$  equals  $P(B)$  times the probability that  $A$  also occurs given  $B$ .

# Applying the Multiplication Rule

## Example 13.5

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**Context:** A 2023 NCAA survey found that betting rates differ by gender among 18–22 year olds. We want to find the probability that a randomly selected person is both male and bets on sports.

- 67% of males bet on sports
- 51% of females bet on sports
- Sample: 50% male, 50% female

**Find:**  $P(\text{male and bets on sports})$

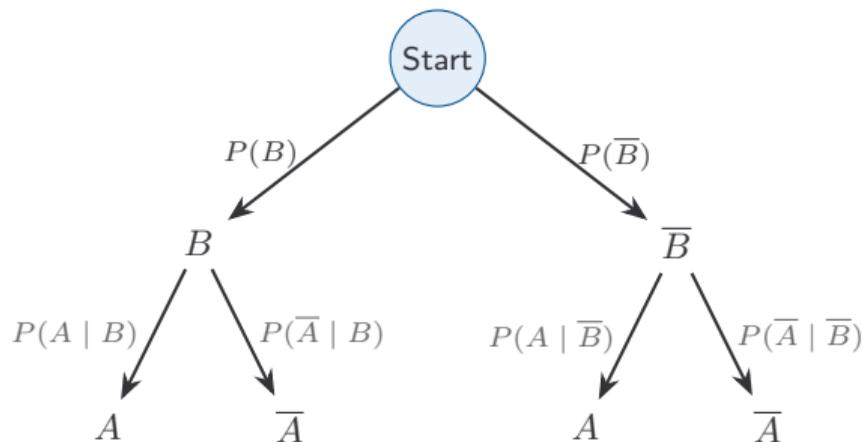


# Tree Diagrams: A Visual Tool for Multi-Stage Problems

## Tree Diagram

A **tree diagram** displays all possible outcomes of a sequence of events, with each branch showing a conditional probability.

**Key Point:** To find the probability of a path:  
**Multiply** the probabilities along the branches.



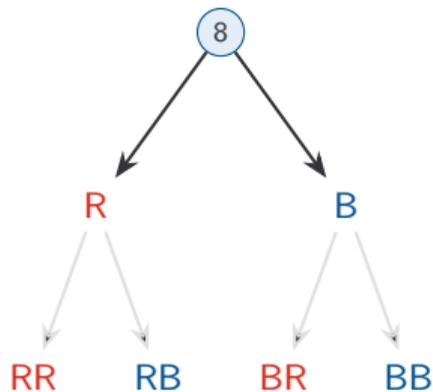
# Tree Diagram for Drawing Without Replacement

Example 13.6

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**Context:** A box has 5 red and 3 blue marbles. You draw two marbles one at a time without putting the first one back.

**Calculate:** Use the tree diagram to find the probabilities below.



(a)  $P(\text{both red})$

(b)  $P(\text{2nd blue} \mid \text{1st red})$

(c)  $P(\text{exactly one red})$

# Multiplication Rule for Three or More Events

## Example 13.7 — Setup

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### Extended Multiplication Rule

For events  $A$ ,  $B$ , and  $C$ :

$$P(A \cap B \cap C) = P(A) \cdot P(B | A) \cdot P(C | A \cap B)$$

**Context:** Based on Western University's Schulich School of Medicine admissions process. Three sequential stages must all be passed:

- Screening: 45% of applicants pass
- Interview: 40% of those screened pass
- Final Review: 30% of those interviewed are admitted

**Find:** What is the probability that a random applicant makes it through all three stages?

# Multiplication Rule for Three or More Events

## Example 13.7 — Calculation

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**Context:** Screening (45%), Interview (40%), Final Review (30%).

**Calculate:**  $P(\text{applicant passes all three stages})$



PART 3

# Independence

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When does knowing  $B$  tell us nothing about  $A$ ?

## Independent Events: Definition

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### Independence

Two events  $A$  and  $B$  are **independent** if

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

- Knowing  $B$  occurred does not change the probability of  $A$ .
- The general multiplication rule simplifies: no conditional probability needed.

# Multiplication Rule for Independent Events

## Example 13.9

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Roll a third die. Let  $a$  be the event of rolling a two, and let  $b$  be the event of rolling an even number. Determine if these two events are independent

If  $A$  and  $B$  are **independent**, and  $B$  is an event with non-zero probability, then

$$P(A | B) = P(A).$$

In other words, knowing that  $B$  occurred does not change the probability of  $A$ .

# Testing for Independence

## Example 13.10 — Setup

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**Context:** Does having a meal plan make students more likely to eat breakfast? A survey of 200 university students found:

- 120 have a meal plan; of these, 90 eat breakfast daily.
- 80 do not have a meal plan; of these, 30 eat breakfast daily.

**Test:** Are having a meal plan ( $M$ ) and eating breakfast daily ( $B$ ) independent?

To test independence, check either:

- Is  $P(B \mid M) = P(B)$ ?
- Or: Is  $P(B \cap M) = P(B) \cdot P(M)$ ?



## A Subtle Difference in Conditioning

Example 13.11

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**Context:** A household adopts two dogs. Each dog is equally likely to be male (M) or female (F), independently.

The sample space is:  $\{MM, MF, FM, FF\}$ , each with probability  $\frac{1}{4}$ .

(a)  $P(\text{both male} \mid \text{first is male})$


(b)  $P(\text{both male} \mid \text{at least one male})$




**Key Point:** The answers differ because the conditions restrict the sample space differently. “The first is male” leaves 2 outcomes; “at least one is male” leaves 3.



## The Gambler's Fallacy

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At a roulette table, the last 7 spins landed on red. Should you bet on black?

 **Caution: The Gambler's Fallacy:** Believing past independent events affect future probabilities.

Each spin is **independent**. The wheel has no memory.  $P(\text{Black}) = \frac{18}{38}$  regardless of history.

PART 4

# Law of Total Probability & Bayes' Rule

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Reversing the direction of conditional probability

# Partitioning the Sample Space

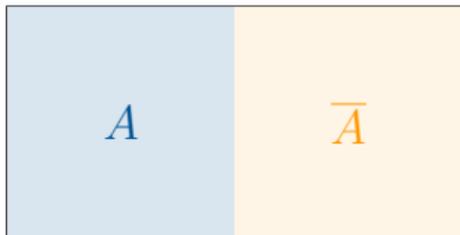
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## Partition

Events  $A_1, A_2, \dots, A_k$  form a **partition** of the sample space  $S$  if:

1. They are **mutually exclusive**: no two can occur at the same time.
2. They are **exhaustive**: together they cover all of  $S$ .

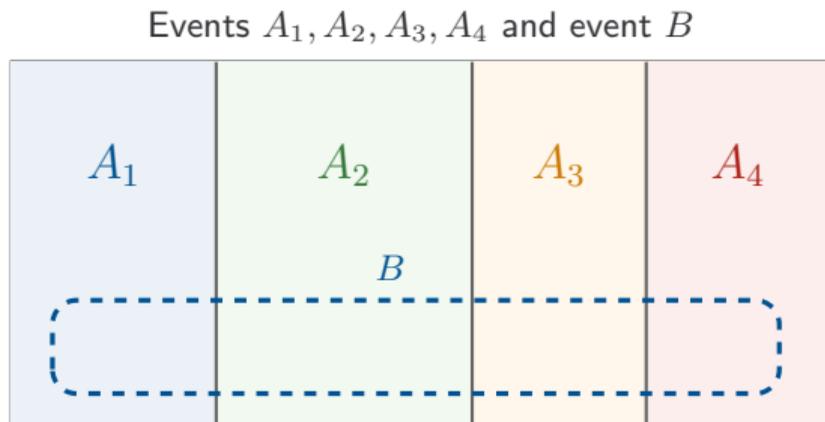
Simplest partition:  $A$  and  $\bar{A}$



**Key Point:** Every outcome in  $S$  falls into exactly one piece of the partition. No gaps, no overlaps.

## A More Complex Partition

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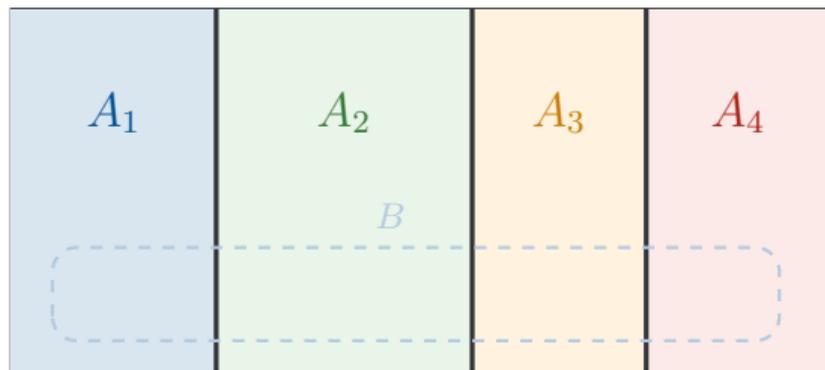


The events  $A_1, A_2, A_3, A_4$  divide the sample space into regions. Event  $B$  can overlap with several of them.

## The Partition: Mutually Exclusive and Exhaustive

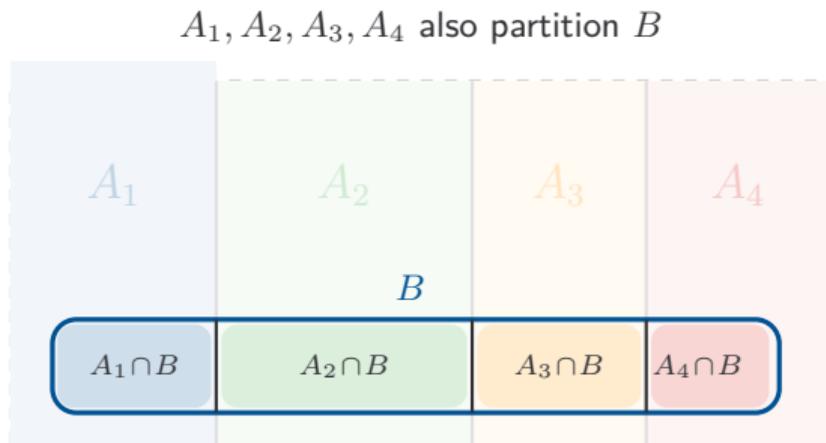
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$A_1, A_2, A_3, A_4$  partition the sample space  $S$



**Key Point:** The  $A_i$  are mutually exclusive (no overlaps) and exhaustive (no gaps). Every outcome in  $S$  belongs to exactly one  $A_i$ .

## The Partition Also Splits $B$



**Key Point:** The partition of  $S$  creates a partition of  $B$ : every outcome in  $B$  falls into exactly one  $A_i \cap B$ . This is the foundation of the Law of Total Probability.



# Law of Total Probability

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## Law of Total Probability

If  $A_1, A_2, \dots, A_k$  partition  $S$ , then

$$P(B) = \sum_{i=1}^k P(B | A_i) P(A_i)$$

for any event  $B$ .

## Applying the Law of Total Probability

Example 13.7 — Extended

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**Context:** Returning to the bookstore laptop problem. A laptop is selected at random and found to be defective. Which supplier is it most likely from?

Recall:

Supplier	Share	Defect Rate
TechCo	40%	2%
BuildRight	25%	5%
QuickChip	20%	4%
ValuePC	15%	10%

**Find:**  $P(\text{ValuePC} \mid \text{Defective})$



## Reversing Conditional Probabilities

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We often know  $P(B | A)$  but need  $P(A | B)$ .

**Context:** A medical test for a rare disease:

- The test correctly detects 95% of people who *have* the disease.
- The test correctly clears 90% of people who *do not* have it.

You test positive. What is the probability you actually have the disease? We know  $P(+ | D)$ , but we need  $P(D | +)$ .

 **Key Point:** Bayes' Rule lets us “flip” the direction of a conditional probability.

This is essential for:

- Medical diagnosis (COVID tests)?
- Quality control

# Bayes' Rule

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## Bayes' Rule

For events  $A$  and  $B$  with  $P(A) > 0$ ,  $P(B) > 0$ :

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

Where does  $P(B)$  come from? Use the Law of Total Probability:

$$P(B) = P(B | A) P(A) + P(B | \bar{A}) P(\bar{A})$$

 **Key Point: Numerator:** Probability of pathway  $A \rightarrow B$ .

**Denominator:** Total probability of  $B$  from all pathways.

What fraction of all paths to  $B$  came through  $A$ ?

# COVID-19 Rapid Antigen Tests

## Example 13.12 — Setup

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**Context:** A rapid antigen test is used to screen for COVID-19. Based on a large study (117,372 samples):

- 5% of the population currently has COVID.
- The test correctly identifies 73% of people who have COVID (true positive rate).
- Among people who do *not* have COVID, 0.7% incorrectly test positive (false positive rate).

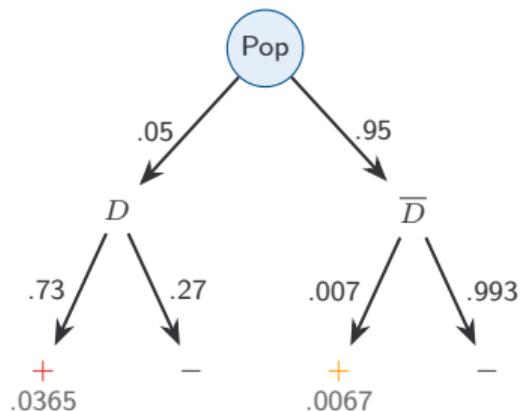
**Find:** A person tests positive. What is the probability they actually have COVID?

 **Caution:** Even a test with a low false positive rate can produce many false positives when the disease is rare.

# COVID-19 Rapid Antigen Tests

## Example 13.12 — Tree Diagram

**Context:** 5% have COVID. The test correctly identifies 73% of those with COVID and incorrectly flags 0.7% of those without.



Two pathways lead to a positive test:

- **Has COVID and tests positive:**  
 $0.05 \times 0.73 = 0.0365$
- **No COVID but tests positive:**  
 $0.95 \times 0.007 = 0.0067$

**Find:** What fraction of all positive tests are true positives?

# COVID-19 Rapid Antigen Tests

## Example 13.12 — Calculation

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**Calculate:** Use Bayes' Rule to find  $P(\text{COVID} \mid +)$ .



# AI Writing Detection (Turnitin)

## Example 13.13 — Setup

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**Context:** Universities use Turnitin to flag student papers that may have been written by AI. A student's paper is flagged. Should the instructor assume it was AI-written?

Here is what we know about the detector's accuracy:

- It correctly flags 85% of papers that *are* AI-written.
- It incorrectly flags 1% of papers that are *human*-written.
- Suppose 10% of all submissions contain significant AI content.

**Find:** If a paper is flagged, what is the probability it was actually AI-written?

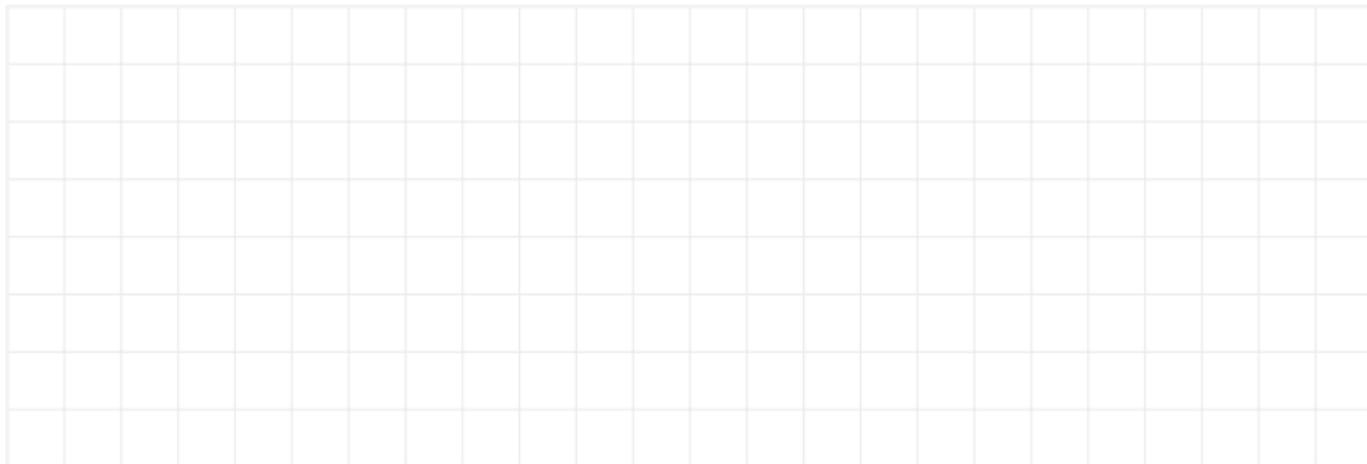
# AI Writing Detection (Turnitin)

## Example 13.13 — Calculation

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**Context:** 85% detection rate, 1% false positive rate, 10% base rate of AI use.

**Calculate:**  $P(\text{AI-written} \mid \text{flagged})$



# When to Use Bayes' Rule

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 **Key Point:** Use Bayes' Rule when you need to **reverse** a conditional probability:

- You know  $P(\text{result} \mid \text{condition})$
- You need  $P(\text{condition} \mid \text{result})$

## Bayes' Rule Steps

1. Identify:  $P(B \mid A)$ ,  $P(A)$
2. Need:  $P(A \mid B)$
3. Find  $P(B)$  via Total Probability
4. Apply:  $P(A \mid B) = \frac{P(B \mid A) \cdot P(A)}{P(B)}$

PART 5

# Common Mistakes

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Three errors to watch for in probability reasoning

## Confusing $P(A | B)$ with $P(B | A)$

### Common Mistake 1

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**⚠ Caution:** The direction of the conditional matters.  $P(A | B) \neq P(B | A)$  in general.

“The test is 95% accurate for people with the disease, so if I test positive, there is a 95% chance I have the disease.”

This confuses  $P(+ | D)$  with  $P(D | +)$ .

#### Quick example:

- $P(\text{wet} | \text{rain}) \neq P(\text{rain} | \text{wet})$
- $P(\text{cough} | \text{flu}) \neq P(\text{flu} | \text{cough})$
- $P(\text{bark} | \text{dog}) \neq P(\text{dog} | \text{bark})$

**🔑 Key Point:** In the COVID test example:  $P(+ | D) = 73\%$  but  $P(D | +) \approx 85\%$ . These are different questions with different answers.

# Assuming Independence Without Checking

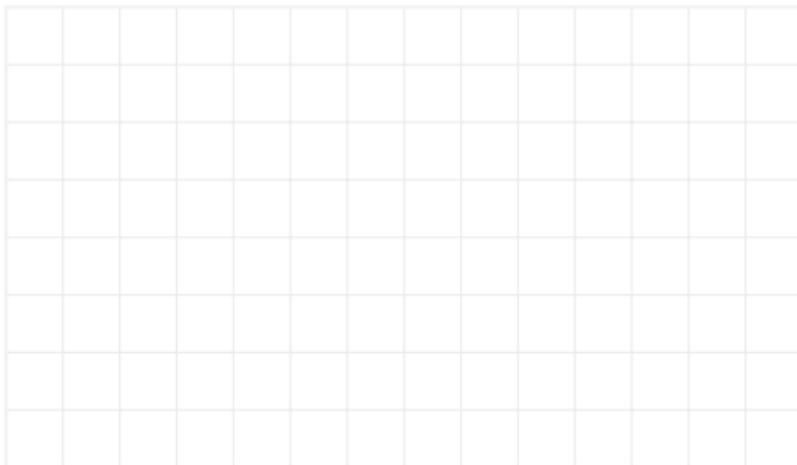
## Common Mistake 2

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NCAA survey data on 3,527 young adults:

	<b>Bets</b>	<b>Not</b>	<b>Tot</b>
Male	1,182	582	1,764
Female	899	864	1,763
<b>Total</b>	2,081	1,446	3,527

Are gender (M) and betting (B) independent?





## Summary: Avoiding Common Probability Mistakes

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1. **Direction matters.**  $P(A | B) \neq P(B | A)$  in general.  
 $P(\text{rain} | \text{clouds}) \neq P(\text{clouds} | \text{rain})$ .
2. **Do not assume independence.** Always verify mathematically or from context.  
Events that seem unrelated may still be dependent.
3. **Mutually exclusive  $\neq$  independent.** They are opposite concepts.  
Mutually exclusive events (with  $P > 0$ ) are always dependent.

CHAPTER 13

# Summary

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Key concepts and formulas from this chapter

## Chapter 13 Summary

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### ■ Conditional Probability

- $P(A | B) = \frac{P(A \cap B)}{P(B)}$
- Restricts sample space to  $B$
- Direction matters.

### ■ Multiplication Rule

- $P(A \cap B) = P(B) \cdot P(A | B)$
- Extends to 3+ events
- Tree diagrams help visualize

### ■ Independence

- $P(A | B) = P(A)$
- If independent:  
 $P(A \cap B) = P(A) \cdot P(B)$
- $\neq$  mutually exclusive.

### ■ Law of Total Probability

- $P(B) = \sum P(B | A_i) P(A_i)$
- Sums over all pathways to  $B$

### ■ Bayes' Rule

- $P(A | B) = \frac{P(B|A) \cdot P(A)}{P(B)}$
- Reverses conditional direction

# Looking Back and Looking Ahead

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 **Key Point:** Chapter 13 introduced tools for reasoning about related events: how knowing one thing changes the probability of another.

## Key Skills:

- Compute conditional probabilities
- Use multiplication rule and trees
- Test for independence
- Apply Bayes' Rule

## Real-World Applications:

- Medical testing (COVID-19 rapid tests)
- AI writing detection (Turnitin)
- Sports betting and risk assessment
- Airline on-time performance

PRACTICE

# Problems

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Test your understanding of conditional probability, independence, and Bayes' Rule

## Practice Problem 1: Conditional Probability

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**Context:** A survey of 500 employees at a company:

	Full-Time	Part-Time	Total
Remote	120	80	200
In-Office	210	90	300
Total	330	170	500

- (a) Find  $P(\text{Remote})$ .
- (b) Find  $P(\text{Remote} \mid \text{Part-Time})$ .
- (c) Find  $P(\text{Full-Time} \mid \text{Remote})$ .
- (d) Are working remotely and being full-time independent? Justify.

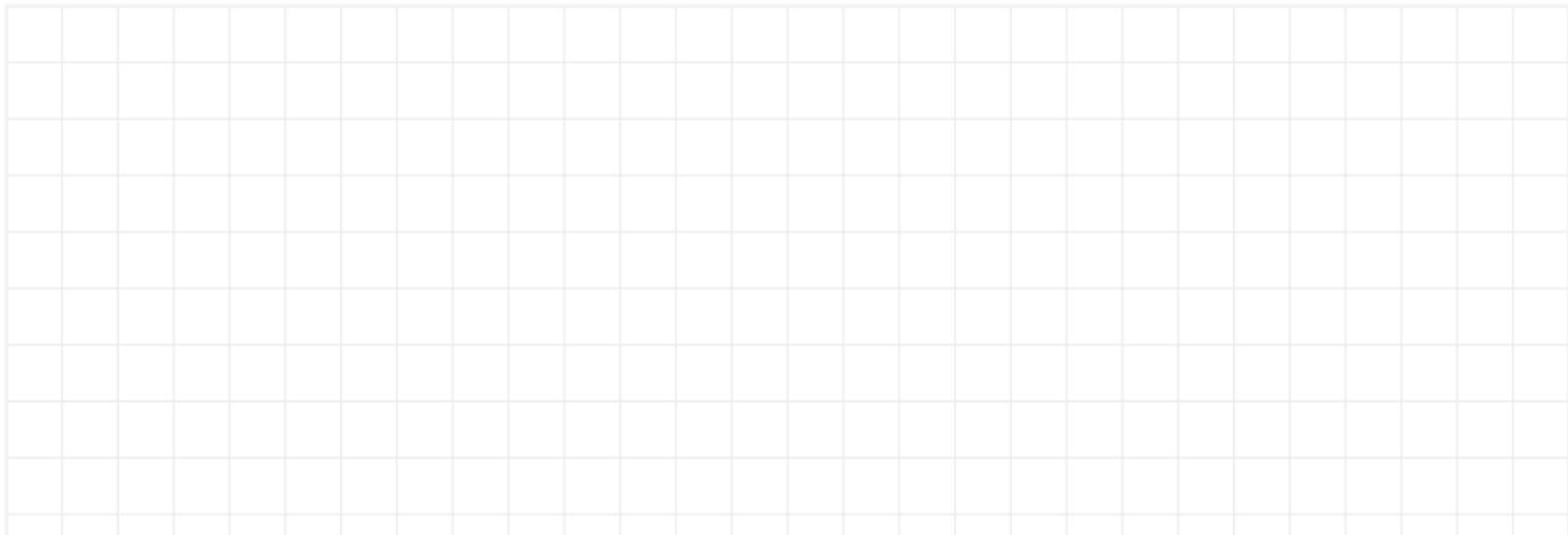


## Practice Problem 2: Multiplication Rule and Trees

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**Context:** A bag contains 4 red, 3 green, and 2 blue marbles. You draw two marbles without replacement.

- (a) Draw a tree diagram for the first two draws (group by color).
- (b) Find  $P(\text{both marbles are red})$ .
- (c) Find  $P(\text{second marble is green} \mid \text{first marble is red})$ .
- (d) Find  $P(\text{at least one marble is blue})$ .





## Practice Problem 4: Bayes' Rule

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**Context:** Three major U.S. airlines have the following market shares and on-time arrival rates (BTS/OAG, 2023):

- Delta: 35% of flights, 83% on-time
  - United: 30% of flights, 79% on-time
  - American: 35% of flights, 79% on-time
- (a) If a random domestic flight is selected, what is the probability it arrives on time?
- (b) If a flight is late, what is the probability it was a Delta flight?
- (c) If a flight is late, which airline is it most likely from?





