

COMSC 2043

Counting

Student Workbook

Jeremy Evert
Southwestern Oklahoma State University

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Chapter 1

Counting: Combinatorial Foundations of Computer Science

How Chapter 6 Fits into the Big Picture

In the grand journey of discrete mathematics, **Chapter 6: Counting** marks a turning point. Up to this point, we have focused on the language of logic (Chapter 1), the structures that give shape to mathematical reasoning such as sets, functions, and sequences (Chapter 2), and the design of step-by-step procedures through algorithms (Chapter 3). We have also examined how numbers behave and how to prove properties about them, both through number theory and by building proofs inductively and recursively (Chapters 4 and 5).

Now, the focus shifts from *why* things are true to *how many* ways they can be true. Counting gives us the ability to measure the size of possibilities—whether it’s the number of valid passwords, the number of paths through a network, or the number of ways to schedule tasks on a processor. In computer science, these aren’t abstract curiosities: they are the foundation of data analysis, probability, algorithmic efficiency, and cryptographic security.

Overview of Chapter 6

This chapter introduces the fundamental principles of **combinatorics**—the art and science of counting. Each section develops a new tool in the combinatorial toolkit:

6.1 The Basics of Counting introduces the rule of sum and the rule of product, the two fundamental ideas that allow us to count complex structures by breaking them into simpler cases.

6.2 The Pigeonhole Principle formalizes an intuitive idea: if you have more objects than containers, at least one container must hold more than one object. This principle turns up everywhere—from hashing algorithms to error detection and compression schemes.

6.3 Permutations and Combinations explores ordered and unordered selections—how to count arrangements of items with or without repetition. These are the cornerstones of combinatorial reasoning and underpin probability theory and algorithmic enumeration.

6.4 Binomial Coefficients and Identities connects counting to algebra through Pascal’s triangle and the binomial theorem, showing how algebraic expressions encode combinatorial ideas.

6.5 Generalized Permutations and Combinations extends our tools to more complex scenarios: multisets, repeated elements, and the counting of indistinguishable objects.

6.6 Generating Permutations and Combinations turns theory into practice. Here we explore how to systematically produce combinations and permutations using algorithmic logic—a perfect bridge between mathematics and computer science.

Each topic builds naturally on the last, leading toward a robust framework for reasoning about *how many* ways something can happen—a question that lies at the heart of computational problem solving.

Counting in the Broader Context of Computer Science

Counting is more than arithmetic—it is **computation in miniature**. Every time a program loops through possibilities, a scheduler distributes resources, or a cryptographer designs a key system, counting quietly determines what’s possible and what’s practical. The efficiency of algorithms often depends on how many steps or arrangements exist; probability models rely on counting possible outcomes; and even machine learning models depend on combinatorial structures when exploring feature combinations or optimization paths.

In short, the study of counting provides a bridge between abstract reasoning and algorithmic design. It transforms intuition into strategy, helping computer scientists predict complexity, measure growth, and understand limits.

By mastering this chapter, you are learning to see the invisible arithmetic beneath every decision tree, database index, and combinational circuit—a skill as essential to the digital age as logic and code themselves.

Chapter 2

Section 6.1 — The Basics of Counting

2.1 Introduction

Suppose that a password on a computer system consists of six, seven, or eight characters. Each of these characters must be a digit or a letter of the alphabet, and each password must contain at least one digit. How many such passwords are there?

Questions like this form the beating heart of discrete mathematics: *How many ways can something happen?* Counting allows us to measure complexity, probability, and possibility. In computer science, every algorithm that loops, branches, or searches is secretly counting something.

2.2 Basic Counting Principles

These two ideas—sum and product—form the foundation of all combinatorial reasoning.

The Sum Rule If a task can be done in n_1 ways *or* n_2 ways (but not both), then it can be done in $n_1 + n_2$ ways.

The Product Rule If a task can be broken into two independent subtasks—first done in n_1 ways, then in n_2 ways—then the whole procedure can be done in $n_1 \times n_2$ ways.

2.3 Worked Examples

Example 2.1 (Assigning Offices). A new company with two employees, Sanchez and Patel, rents a floor with 12 offices. How many ways are there to assign different offices to these two employees?

Solution. 12 choices for Sanchez, then 11 remaining for Patel. By the product rule: $12 \times 11 = 132$ possibilities.

Example 2.2 (Labeling Auditorium Chairs). Each seat is labeled with one uppercase English letter followed by a positive integer not exceeding 100. How many unique chair labels exist?

Solution. 26 possible letters \times 100 integers = 2600 labels.

Example 2.3 (Counting Ports in a Data Center). There are 32 computers, each with 24 ports. How many total ports exist?

Solution. $32 \times 24 = 768$ ports.

Example 2.4 (Challenging A: License Plate Combinations). A region issues license plates with three uppercase letters followed by three digits (e.g., ABC123). How many distinct plates are possible if repetition is allowed?

Solution. $26^3 \times 10^3 = 175,760,000$ possible plates.

If letters and digits cannot repeat, the count becomes $26 \times 25 \times 24 \times 10 \times 9 \times 8 = 112,320,000$.

Example 2.5 (Challenging B: Passwords with a Digit Requirement). A password must be 6, 7, or 8 characters long, each either a letter or digit, and must contain at least one digit. How many possible passwords exist?

Solution. For each length n , count all strings (36^n) and subtract those with only letters (26^n):

$$P_n = 36^n - 26^n$$

Summing across the three allowed lengths:

$$\text{Total} = P_6 + P_7 + P_8 = (36^6 - 26^6) + (36^7 - 26^7) + (36^8 - 26^8)$$

2.4 Python Demonstrations

Let's translate these counting ideas into executable logic. Below is Python code that mirrors the reasoning in these examples. Each section prints both the symbolic reasoning and the computed value. Students are encouraged to modify the numbers and observe how the results change.

Listing 2.1: Demonstrating the Product and Sum Rules in Python

```
1 """
```



```
2 COMSC_2043_Counting_Demonstrations
3 Author: Jeremy Evert
4 This script demonstrates the basic principles of counting using Python.
5 """
6
7 from math import comb, perm
8
9 def example_offices():
10     n1, n2 = 12, 11
11     total = n1 * n2
12     print(f"Example 1: Assigning offices\n12*11={total}\n")
13
14 def example_chairs():
15     total = 26 * 100
16     print(f"Example 2: Chair labeling\n26*100={total}\n")
17
18 def example_ports():
19     total = 32 * 24
20     print(f"Example 3: Data center ports\n32*24={total}\n")
21
22 def example_license_plates():
23     allow_repeat = 26**3 * 10**3
24     no_repeat = (26*25*24) * (10*9*8)
25     print("Example 4: License plates")
26     print(f"    With repetition: {allow_repeat:,}")
27     print(f"    Without repetition: {no_repeat:,}\n")
28
29 def example_passwords():
30     # Helper function: A^n - B^n
31     def count_valid(length):
32         return 36**length - 26**length
33
34     total = sum(count_valid(n) for n in (6, 7, 8))
35     print("Example 5: Passwords with at least one digit")
36     print(f"    Total valid passwords: {total:,}\n")
37
38 def main():
39     print("=== Counting Demonstrations ===\n")
```

```
40     example_offices()
41     example_chairs()
42     example_ports()
43     example_license_plates()
44     example_passwords()
45     print("Done!")
46
47 if __name__ == "__main__":
48     main()
```

Discussion and Reflection

Each of these problems could be solved by intuition, but writing code forces precision. Python doesn't "believe" in the product rule—it *performs* it. This helps students see that the abstract logic of counting translates directly into computation.

Encourage students to:

- Change numbers and verify the rules still hold.
- Add print statements to trace intermediate steps.
- Reflect on where independence between tasks exists—and where it doesn't.

Key Takeaway. Every counting problem begins with one question: Are the choices independent or exclusive? From that single distinction, the sum and product rules unfold.

Next Steps

In the next section we will apply these rules to cases where choices overlap or restrict one another—leading to the powerful and deceptively simple **Pigeonhole Principle**.

Chapter 3

Section 6.1 — Extended Product Rule

3.1 The Extended Product Rule

The basic product rule can be generalized to any number of sequential tasks. Suppose a procedure consists of tasks T_1, T_2, \dots, T_m , where each task T_i can be completed in n_i ways, independent of how the earlier tasks were done. Then the entire procedure can be performed in

$$n_1 \times n_2 \times \cdots \times n_m.$$

This generalization is called the **Extended Product Rule**. It can even be proven formally using mathematical induction on the number of tasks—an idea that connects beautifully to Chapter 5 on induction and recursion.

In essence, when tasks are independent and sequential, the total number of possible outcomes is the product of the possibilities for each step.

3.2 Worked Examples

The following examples illustrate how the extended product rule applies in both mathematics and computer science.

Example 3.1 (Bit Strings of Length Seven). How many different bit strings of length seven are there?

Solution. Each of the seven bits can be chosen in two ways—either 0 or 1. By the

product rule, the total number of bit strings is:

$$2^7 = 128.$$

Example 3.2 (License Plates). How many different license plates can be made if each contains three uppercase English letters followed by three digits?

Solution. There are 26 choices for each of the three letters and 10 choices for each of the three digits:

$$26^3 \times 10^3 = 17,576,000.$$

So there are 17,576,000 possible license plates.

Example 3.3 (Counting Functions). How many functions exist from a set with m elements to a set with n elements?

Solution. Each element of the domain can be mapped to any of the n elements in the codomain. Thus, by the product rule:

$$n^m \text{ functions.}$$

For instance, there are $5^3 = 125$ functions from a three-element set to a five-element set.

Example 3.4 (Counting One-to-One Functions). How many one-to-one (injective) functions exist from a set with m elements to a set with n elements?

Solution. If $m > n$, none exist. When $m \leq n$, the first element can map to any of n values, the next to $(n - 1)$ remaining values, and so on, giving:

$$n(n - 1)(n - 2) \cdots (n - m + 1).$$

For example, from a three-element set to a five-element set:

$$5 \times 4 \times 3 = 60.$$

3.3 Apply It Yourself

Try these practice problems to reinforce the rule. Each one extends the pattern of independent sequential choices.

Warm-Up Problem

How many different bit strings of length eight are there?

Solution. Each bit can be either 0 or 1, so there are $2^8 = 256$ possible bit strings.

Practice Problem 1 — Easier

How many license plates can be made if each plate has two uppercase English letters followed by two digits?

Solution. Each of the two letters has 26 possibilities, and each digit has 10:

$$26^2 \times 10^2 = 67,600.$$

Thus, there are 67,600 unique plates.

Practice Problem 2 — Moderate

How many one-to-one functions are there from a set with four elements to a set with six elements?

Solution.

$$6 \times 5 \times 4 \times 3 = 360.$$

There are 360 one-to-one functions.

Practice Problem 3 — Challenging

How many functions exist from a set with $m = 10$ elements to a set with $n = 3$ elements? How many of these are one-to-one?

Solution.

- Total functions: $3^{10} = 59,049$.
- One-to-one functions: none, since $m > n$.

Practice Problem 4 — Stretch Challenge

A company issues employee IDs consisting of two uppercase letters, followed by one of five department codes (A–E), followed by a 3-digit number (000–999). How many possible IDs can be created?

Solution.

$$26^2 \times 5 \times 1000 = 3,380,000.$$

So there are 3,380,000 unique ID numbers possible.

3.4 Python Demonstrations

Let's express the same reasoning in Python. The following script illustrates how the extended product rule scales as the number of independent tasks grows.

Listing 3.1: Extended Product Rule Examples

```

1  """
2  COMSC_2043_-_Extended_Product_Rule_Demonstrations
3  Author:_Jeremy_Evert
4  """
5
6  from math import prod
7
8  def bit_strings(n):
9      return 2 ** n
10
11 def license_plates(letters, digits):
12     return 26 ** letters * 10 ** digits
13
14 def functions(m, n):
15     return n ** m
16
17 def one_to_one_functions(m, n):
18     if m > n:
19         return 0
20     choices = [n - i for i in range(m)]
21     return prod(choices)
22
23 def main():
24     print("===_Extended_Product_Rule_===\n")
25     print(f"Bit_strings_of_length_7:_{bit_strings(7)}")
26     print(f"License_plates_(3_letters,_3_digits):_{license_plates(3,3):,}")
27     print(f"Functions_from_3->5:_{functions(3,5)}")
28     print(f"One-to-one_from_3->5:_{one_to_one_functions(3,5)}")

```

```
29     print(f"Warm-up_{8-bit_strings}:{bit_strings(8)}")
30     print(f"Practice_1_{2_letters,2_digits}:{license_plates(2,2):,}")
31     print(f"Practice_2_{one-to-one_4->6}:{one_to_one_functions(4,6)}")
32     print(f"Practice_3_{functions_10->3}:{functions(10,3)}|_one-to-one:_{one_to_one_functions(10,3)}")
33     print(f"Practice_4_{employee_IDs}:{26**2*5*1000:,}")
34     print("\nDone!")
35
36 if __name__ == "__main__":
37     main()
```

Reflection

The extended product rule shows how independence and sequence multiply possibilities. The pattern $n_1 n_2 \cdots n_m$ appears everywhere—from network addressing and password generation to database schema design and combinatorial search. Every time we build loops or nested conditionals in code, we are implicitly applying this same principle.

Key Takeaway. Each new layer of choice multiplies the universe of outcomes. Recognizing independence is the key to mastering both counting and algorithmic reasoning.

Next Steps

In the next chapter, we'll watch the product rule come alive in real systems— from telephone numbers and nested loops to subsets and combinatorial trees— as we explore **The Product Rule in Action**.

Chapter 4

Section 6.1 — Product Rule in Action

4.1 Real-World Applications of the Product Rule

Counting problems appear everywhere—from numbering systems to programming loops and data structures. The product rule quietly powers many of the systems we rely on every day. Let’s see how it works in the wild.

4.2 Example 1 — The North American Telephone Numbering Plan

The North American Numbering Plan (NANP) defines 10-digit phone numbers, split into a three-digit *area code*, a three-digit *office code*, and a four-digit *station code*.

- N : any digit from 2–9
- Y : either 0 or 1
- X : any digit from 0–9

Old Plan (NYX–NNX–XXXX):

Area codes: $8 \times 2 \times 10 = 160$,
Office codes: $8 \times 8 \times 10 = 640$,
Station codes: $10^4 = 10,000$.

Total: $160 \times 640 \times 10,000 = 1.024 \times 10^9$ phone numbers.

New Plan (NXX–NXX–XXXX):

Area codes: $8 \times 10 \times 10 = 800$,

Office codes: $8 \times 10 \times 10 = 800$,

Station codes: $10^4 = 10,000$.

Total: $800 \times 800 \times 10,000 = 6.4 \times 10^9$ phone numbers.

Solution. The new plan increases capacity by a factor of about six. A small rule change leads to billions more combinations—combinatorics in action.

Practice Problems

1. Easier. A small town uses 7-digit numbers in the format NXX–XXX, where N is from 2–9 and X is from 0–9. How many phone numbers can be issued?

Solution.

$$8 \times 10^5 = 8,000,000.$$

2. Moderate. If the first two digits must be odd (1, 3, 5, 7, 9), how many numbers exist?

Solution.

$$5 \times 5 \times 10^5 = 2,500,000.$$

3. Stretch. If each number begins with one of 26 region letters (A–Z), how many distinct identifiers exist?

Solution.

$$26 \times 8,000,000 = 208,000,000.$$

4.3 Example 2 — Counting Nested Loops

Programmers apply the product rule constantly, often without realizing it. Consider the pseudocode:

```
1 k = 0
2 for i1 in range(1, n1+1):
3     for i2 in range(1, n2+1):
4         ...
5         for im in range(1, nm+1):
6             k += 1
```

Each loop represents one “task” T_i with n_i iterations. By the product rule, the innermost statement executes:

$$n_1 \times n_2 \times \cdots \times n_m$$

times, so the final value of k is exactly that product.

Practice Problems

1. Easier. If the outer loop runs 5 times and the inner loop 4 times, how many increments occur?

$$5 \times 4 = 20.$$

2. Moderate. For three nested loops running 3, 4, and 2 times respectively:

$$3 \times 4 \times 2 = 24.$$

3. Stretch. Write a Python function that computes k automatically for any sequence (n_1, n_2, \dots, n_m) .

```

1 from math import prod
2 def nested_loops(*sizes):
3     return prod(sizes)
4 print(nested_loops(3, 4, 2)) # Output: 24

```

4.4 Example 3 — Counting Subsets of a Finite Set

Use the product rule to show that a set S with $|S| = n$ has 2^n subsets.

Solution. For each element of S , we make two independent choices:

1. include it in the subset, or
2. exclude it.

Each of the n choices has two outcomes, producing:

$$2^n \text{ total subsets.}$$

Each subset corresponds to a unique bit string of length n —a binary fingerprint of membership.

Practice Problems

1. **Easier.** How many subsets does $\{a, b, c\}$ have?

$$2^3 = 8.$$

2. **Moderate.** If $|S| = 10$, how many subsets contain at least one element?

$$2^{10} - 1 = 1023.$$

3. **Stretch.** How many subsets of a set with n elements have exactly k members?

$$\binom{n}{k}.$$

4.5 Python Demonstrations

Listing 4.1: Counting with Loops and Sets

```
1 from itertools import chain, combinations
2 from math import prod
3
4 # Telephone numbering comparison
5 def phone_numbers(area, office, station):
6     return area * office * station
7
8 print("Old_plan:", phone_numbers(160, 640, 10_000))
9 print("New_plan:", phone_numbers(800, 800, 10_000))
10
11 # Nested loop counter
12 def nested_count(*sizes):
13     return prod(sizes)
14
15 print("Nested_loop_3x4x2:", nested_count(3,4,2))
16
17 # Subsets demonstration
18 S = {'a', 'b', 'c'}
19 power_set = list(chain.from_iterable(combinations(S, r) for r in range(len(S)+1)))
20 print("Subsets_of_{a,b,c}:", power_set)
21 print("Count:", len(power_set))
```

Reflection

The product rule connects telephone networks, program loops, and the logic of sets. Each system multiplies independent possibilities, revealing how small rules create enormous complexity.

In computer science, this principle hides in plain sight: every nested loop, every binary decision, every permutation generator is a quiet tribute to the power of multiplication.

Next Steps

In the next chapter, we turn to the other side of counting— where overlap, exclusion, and repetition break independence— and discover the elegance of the **Sum and Subtraction Rules**.

Chapter 5

Section 6.2 — DNA, Genomes, and the Sum Rule

5.1 DNA and Combinatorics

DNA (deoxyribonucleic acid) and RNA (ribonucleic acid) encode the very instructions of life. Each strand of DNA consists of *nucleotides*, with four possible bases:

A (adenine), C (cytosine), G (guanine), T (thymine).

In RNA, thymine (T) is replaced by uracil (U). These bases form long chains whose order determines genes and, ultimately, the proteins that define an organism.

From a combinatorial perspective, each base position can be filled in one of four ways. Hence, a DNA strand with n bases corresponds to 4^n possible sequences — the **product rule** in biological disguise.

—

Example 5.1 (Encoding Amino Acids). Proteins are built from amino acids. Humans use 22 essential amino acids, and each amino acid is encoded by a *codon* — a short sequence of bases.

Reasoning:

- One base: $4^1 = 4$ possibilities (not enough).
- Two bases: $4^2 = 16 < 22$ (still too few).
- Three bases: $4^3 = 64$ (more than enough).

Therefore, a codon must consist of three bases, providing 64 distinct triplets to encode 22 amino acids. This redundancy helps prevent catastrophic errors from single-base mutations.

Solution. Nature, like a clever coder, includes redundancy and error recovery in its data encoding. The product rule explains why three-base codons are the smallest possible unit that can represent all amino acids.

Example 5.2 (Counting DNA Sequences). DNA length varies dramatically among organisms:

- Simple organisms (e.g., bacteria): 10^5 – 10^7 bases.
- Complex organisms (e.g., mammals): 10^8 – 10^{10} bases.

By the product rule, a DNA molecule of length n has 4^n possible base sequences:

$$4^{10^5} \quad \text{for bacteria,} \quad 4^{10^8} \quad \text{for mammals.}$$

These quantities are astronomically large, illustrating why genetic diversity is effectively limitless.

Solution. Even though DNA uses just four symbols, its combinatorial potential is enormous. This exponential explosion of possible sequences ensures endless biological variety — one reason every living thing is unique.

Practice Problems — DNA and Counting

1. **Easier:** How many distinct RNA sequences of length four are possible?

Solution: Each position has four choices (A, C, G, U):

$$4^4 = 256 \text{ possible sequences.}$$

—

2. Moderate: A gene contains 12 bases. How many unique base sequences can it have?

Solution:

$$4^{12} = 16,777,216.$$

Even a short gene can produce over sixteen million variations.

—

3. Stretch: A virus has a genome with 30,000 bases. Estimate $4^{30,000}$ in powers of ten.

Solution:

$$4^{30,000} = (2^2)^{30,000} = 2^{60,000} \approx 10^{18,000}.$$

That's a one followed by eighteen thousand zeros — a vast genetic universe.

—

5.2 Introducing the Sum Rule

So far, our counting used multiplication — combining independent choices. But what if we have *mutually exclusive* options, and we must choose exactly one? That's where the **sum rule** comes in.

Definition 5.1 (Sum Rule). If a task can be done in n_1 ways or in n_2 ways, where the sets of outcomes are disjoint (no overlap), then the total number of ways is:

$$n_1 + n_2.$$

—

Example 5.3 (Faculty or Student Representative). Suppose either a member of the mathematics faculty or a mathematics major is chosen to serve on a university committee.

Given:

37 faculty members, 83 students.

Since no one is both faculty and student, the total number of possible representatives is:

$$37 + 83 = 120.$$

Solution. The sum rule models disjoint alternatives — you choose one group or the other, not both. In programming, this parallels an **if/else** decision branch.

—

Example 5.4 (Choosing from Multiple Project Lists). A student can pick one project from any of three lists containing 23, 15, and 19 options, respectively. No project appears on more than one list.

Solution:

$$23 + 15 + 19 = 57$$

possible projects in total.

Solution. Each list represents a distinct pool of choices — separate “doors” leading to different sets of outcomes. Adding their sizes counts all unique possibilities.

—

Practice Problems — Applying the Sum Rule

1. Easier: You can adopt either a cat (6 breeds) or a dog (9 breeds) from the shelter. How many total options are there?

Solution:

$$6 + 9 = 15.$$

2. Moderate: A restaurant offers 12 vegetarian entrées and 8 meat dishes. If you can choose only one, how many possible meals are there?

Solution:

$$12 + 8 = 20.$$

—

3. Stretch: A student can take one course chosen from:

- 5 online classes,
- 3 hybrid classes, or
- 4 in-person classes.

No course appears in more than one category. How many total courses are available?

Solution:

$$5 + 3 + 4 = 12.$$

This illustrates the extended sum rule:

$$n_1 + n_2 + \cdots + n_m.$$

5.3 Python Demonstrations

Listing 5.1: DNA and Sum Rule Demonstrations

```

1  """
2  COMSC_2043_DNA_and_Sum_Rule_Examples
3  Author: Jeremy Evert
4  """
5
6  from math import pow
7
8  # DNA combinatorics
9  def dna_sequences(length):
10     """Returns the number of possible DNA (or RNA) sequences of given length."""
11     return int(pow(4, length))
12
13  # Sum rule demonstrations
14  faculty, students = 37, 83
15  project_lists = [23, 15, 19]
16
17  print("=== DNA and Sum Rule Examples ===\n")
18  print(f"DNA sequences of length 4: {dna_sequences(4)}")
19  print(f"DNA sequences of length 12: {dna_sequences(12):,}")
20  print(f"Committee choices (faculty or students): {faculty+students}")
21  print(f"Total projects across lists: {sum(project_lists)}")
22  print("\nDone!")

```

Reflection

The **product rule** multiplies possibilities; the **sum rule** adds alternatives. Together they are the foundation of all combinatorics — and of all computation. Loops represent multiplication (repetition of actions), while conditional branches represent addition (choosing between paths). From DNA sequences to decision trees, counting principles are the quiet mathematics behind every system of order — living or digital.

Chapter 6

Section 6.3 — Permutations and Combinations

6.1 Counting Arrangements and Selections

Two questions shape almost every counting problem:

1. Does *order* matter?
2. Can we *reuse* elements?

If order matters, we count **permutations**. If order does not matter, we count **combinations**. Let's explore both ideas carefully.

—

Example 6.1 (Permutations — When Order Matters). You have three distinct books and want to know how many ways they can be arranged on a shelf.

Reasoning.

- 3 choices for the first position,
- 2 choices for the second,
- 1 choice for the last.

By the product rule:

$$3 \times 2 \times 1 = 6.$$

We call this $3!$ (“three factorial”). In general, $n! = n \times (n - 1) \times \cdots \times 1$.

Solution. If the books are labeled A, B, C, then the six permutations are:

ABC, ACB, BAC, BCA, CAB, CBA.

Factorials grow explosively— $10! = 3,628,800$ —so even small sets lead to huge numbers of arrangements.

—

Example 6.2 (Permutations of Subsets). How many ways can you arrange four of ten distinct students in a line?

Reasoning. You are selecting four *different* people, and order matters:

$$P(10, 4) = \frac{10!}{(10 - 4)!} = \frac{10!}{6!} = 10 \times 9 \times 8 \times 7 = 5040.$$

Solution. $P(n, r)$ counts the number of ways to arrange r objects from n . Python provides `math.perm(n, r)` to compute this directly.

—

6.2 Combinations — When Order Does Not Matter

When order no longer matters, we divide out repeated arrangements that represent the same group.

Example 6.3 (Combinations Formula). From ten students, how many unique groups of four can be formed?

Reasoning.

$$C(10, 4) = \frac{10!}{4! \times (10 - 4)!} = \frac{10!}{4! 6!} = 210.$$

Solution. Use `math.comb(10, 4)` in Python to verify. Combinations appear in team selection, probability, and sampling.

—

Practice Problems — Permutations and Combinations

1. **Easier:** How many ways can five different books be arranged on a shelf?

Solution:

$$5! = 5 \times 4 \times 3 \times 2 \times 1 = 120.$$

—

2. Moderate: From a group of eight students, how many teams of three can be formed?

Solution:

$$C(8, 3) = \frac{8!}{3! 5!} = 56.$$

—

3. Stretch: How many four-digit PINs can be created using digits 0–9 if:

1. digits cannot repeat, and
2. digits may repeat?

Solution:

$$(a) \text{ No repeats: } P(10, 4) = \frac{10!}{6!} = 5040, \quad (b) \text{ Repeats allowed: } 10^4 = 10,000.$$

Repetition multiplies the space of possibilities dramatically.

6.3 Python Demonstrations

Listing 6.1: Exploring permutations and combinations in Python

```
1 """
2 COMSC_2043_-_Section_6.3:_Permutations_&_Combinations
3 Author:_Jeremy_Evert
4 """
5
6 import math
7
8 # Factorials
9 print("5!=", math.factorial(5))
10
11 # Permutations (order matters)
12 print("P(10,4)=", math.perm(10, 4))
13
14 # Combinations (order doesn't matter)
15 print("C(10,4)=", math.comb(10, 4))
16
17 # Comparing repetition
18 no_repeats = math.perm(10, 4)
19 with_repeats = 10**4
20 print(f"No_repeats:_{no_repeats:,}")
21 print(f"With_repeats:_{with_repeats:,}")
```

Reflection

Permutations and combinations form the twin pillars of counting theory. Adding order multiplies possibilities; ignoring it divides them out. These ideas echo through every algorithm, from password generation to genetic code.

$$\text{Permutations: } P(n, r) = \frac{n!}{(n-r)!}, \quad \text{Combinations: } C(n, r) = \frac{n!}{r!(n-r)!}.$$

Understanding the tension between “order” and “group” transforms abstract formulas into tools for real-world reasoning.

Chapter 7

Section 6.1.3 — More Complex Counting Problems

Why this section matters

Up to now, we’ve used the product rule (choices multiply) and the sum rule (disjoint options add). Many real problems mix both ideas and add constraints (like “must begin with a letter” or “must include at least one digit”). Today’s patterns: (i) break by *length* and *first-character rules*, (ii) count with the product rule, and (iii) subtract forbidden cases (complements and inclusion–exclusion).

7.1 Worked Example A — Short variable names with reserved words

Example 7.1. A toy language allows variable names of length 1 or 2. Characters are alphanumeric (letters/digits) but *the first character must be a letter*. Upper/lowercase are treated the same. Additionally, *five specific two-character strings are reserved* and therefore not allowed as variable names. How many valid variable names exist?

Solution. Let V be the number of valid names. Split by length:

Length 1. First character must be a letter \Rightarrow 26 possibilities.

Length 2. First char: 26 choices (letter). Second char: 36 choices (letter or digit). So

$$26 \cdot 36 = 936 \text{ raw two-character strings.}$$

But five of these two-character strings are reserved. Exclude them:

$$V_2 = 936 - 5 = 931.$$

Total. $V = V_1 + V_2 = 26 + 931 = \boxed{957}$.

Sanity checks. (1) The “first is a letter” constraint is applied to both lengths. (2) The -5 only hits two-character names (not the length-1 pool). (3) We didn’t accidentally double-subtract: reserved items are disjoint from the length-1 set.

Python quick check.

```

1 import string
2 letters = string.ascii_uppercase # treat case-insensitive
3 alnum = string.ascii_uppercase + string.digits
4
5 reserved = {"IF", "DO", "TO", "ON", "GO"} # example 5
6 V1 = len(letters)
7 V2 = sum(1 for a in letters for b in alnum if (a+b) not in reserved)
8 print(V1 + V2) # 957

```

7.2 Worked Example B — 6–8 char passwords, at least one digit

Example 7.2. A password is 6, 7, or 8 characters long. Each character is an uppercase letter or a digit. A password is valid only if it includes *at least one digit*. How many valid passwords are there in total?

Solution. For a fixed length n , count all strings over $\{A \dots Z, 0 \dots 9\}$ and subtract the “all letters” strings:

$$P_n = 36^n - 26^n.$$

Sum over $n \in \{6, 7, 8\}$:

$$P = (36^6 - 26^6) + (36^7 - 26^7) + (36^8 - 26^8).$$

Numerically,

$$36^6 - 26^6 = 2,176,782,336 - 308,915,776 = 1,867,866,560,$$

$$36^7 - 26^7 = 78,364,164,096 - 8,031,810,176 = 70,332,353,920,$$

$$36^8 - 26^8 = 2,821,109,907,456 - 208,827,064,576 = 2,612,282,842,880.$$

Therefore

$$P = 2,684,483,063,360.$$

Common gotchas. (1) The complement must match the exact constraint (“at least one digit” \Rightarrow complement is “no digits”). (2) Don’t multiply by 3 lengths; you *add* across mutually exclusive lengths (sum rule). (3) No double-counting across lengths because 6/7/8 are disjoint cases.

Python spot-check.

```
1 def valid(n): return 36**n - 26**n
2 total = sum(valid(n) for n in (6,7,8))
3 print(f"{total:,}") # 2,684,483,063,360
```

7.3 Try It — Easier (solution on the next page)

A promo code is either *four* or *five* characters long. Each character is a letter or digit, and every promo code must contain *at least one letter*. How many promo codes are possible?

Solution (Easier)

Let Q_n be the count for fixed length n . Use a complement again: “at least one letter” \Rightarrow subtract “no letters” (i.e., *all digits*).

$$Q_n = 36^n - 10^n.$$

Sum for $n \in \{4, 5\}$:

$Q = (36^4 - 10^4) + (36^5 - 10^5) = 1,679,616 - 10,000 + 60,466,176 - 100,000 = 62,035,792.$

7.4 Challenge — Harder (solution on the next page)

Passwords are still 6–8 characters using letters/digits, but now add two constraints:

1. The **first character must be a letter**, and
2. The password must contain **at least two digits** overall.

How many valid passwords exist?

Solution (Harder)

Fix a length $n \in \{6, 7, 8\}$. Force the first character to be a letter (26 ways). The remaining $n - 1$ positions each have 36 choices. We need *at least two digits across those $n - 1$ trailing positions*. Count by complement on the tail:

For the last $n - 1$ positions:

$$\#(\text{at least 2 digits}) = 36^{n-1} - \underbrace{26^{n-1}}_{0 \text{ digits}} - \underbrace{\binom{n-1}{1} \cdot 10 \cdot 26^{n-2}}_{\text{exactly 1 digit}}.$$

Multiply by 26 for the first letter:

$$H_n = 26 \left(36^{n-1} - 26^{n-1} - (n-1) \cdot 10 \cdot 26^{n-2} \right).$$

Compute each n and sum:

$$\begin{aligned} H_6 &= 26(36^5 - 26^5 - 5 \cdot 10 \cdot 26^4) = \boxed{669,136,000}, \\ H_7 &= 26(36^6 - 26^6 - 6 \cdot 10 \cdot 26^5) = \boxed{30,029,584,000}, \\ H_8 &= 26(36^7 - 26^7 - 7 \cdot 10 \cdot 26^6) = \boxed{1,266,414,489,600}. \end{aligned}$$

Therefore,

$$\boxed{H = H_6 + H_7 + H_8 = 1,297,113,209,600.}$$

Why inclusion–exclusion? “At least two digits” means we must remove both the 0-digit and 1-digit cases from the tail. Exactly-one-digit strings on the tail: choose the position ($n - 1$ ways), choose the digit (10), and fill others with letters (26^{n-2}).

Python verification.

```

1 def count_at_least_two_with_first_letter(n):
2     from math import comb
3     tail = 36**(n-1) - 26**(n-1) - comb(n-1,1)*10*26**(n-2)
4     return 26 * tail
5
6 print(sum(count_at_least_two_with_first_letter(n) for n in (6,7,8)))
7 # 1,297,113,209,600

```

Chapter 8

Section 6.1.4 — The Subtraction Rule (Avoiding Double Counting)

8.1 Why This Section Matters

By now, we’ve counted outcomes using multiplication (the **Product Rule**) and addition (the **Sum Rule**). But what happens when categories overlap—when some items fit both descriptions?

The **Subtraction Rule** fixes that problem. It corrects for double-counting and lays the foundation for the more powerful Inclusion–Exclusion Principle later.

8.2 Definition: The Subtraction Rule

Definition 8.1 (Subtraction Rule). If one task can be done in n_1 ways or another in n_2 ways, but some outcomes are counted twice because they belong to both sets (there are $n_{1,2}$ overlaps), then the true total is:

$$n_1 + n_2 - n_{1,2}.$$

Subtracting the intersection ensures every unique outcome is counted once and only once.

8.3 Example 1 — Bit Strings with Overlap (Rosen Example 18)

Example 8.1 (8-bit Strings Starting with 1 or Ending with 00). How many bit strings of length 8 start with 1 *or* end with 00?

Setup:

A = strings starting with 1, B = strings ending with 00.

We want:

$$|A \cup B| = |A| + |B| - |A \cap B|.$$

Count each piece.

$$|A| = 2^7 = 128, \quad |B| = 2^6 = 64, \quad |A \cap B| = 2^5 = 32.$$

$$|A \cup B| = 128 + 64 - 32 = 160.$$

Solution. There are 160 valid 8-bit strings. The subtraction removes the 32 strings that satisfy both conditions. Without it, $128 + 64 = 192$ would double-count those overlaps.

—

8.4 Example 2 — Library Database Overlap

Example 8.2 (Students Enrolled in Math or Computer Science). At SWOSU, 130 students take a math course, 85 take a computer science course, and 42 take both. How many students take at least one of the two?

Apply the subtraction rule:

$$130 + 85 - 42 = 173.$$

Solution. 173 unique students are represented. This same logic powers database joins and query optimization: we must subtract the intersection to avoid duplication.

—

Practice Problems — Subtraction in Action

1. **Easier:** How many 5-bit strings start with 0 or end with 1?

Solution:

$$|A| = 2^4 = 16, \quad |B| = 2^4 = 16, \quad |A \cap B| = 2^3 = 8.$$

$$|A \cup B| = 16 + 16 - 8 = 24.$$

Answer: 24 such bit strings.

—

2. Moderate: How many 10-bit strings start with “11” or end with “000”?

Solution:

$$|A| = 2^8 = 256, \quad |B| = 2^7 = 128, \quad |A \cap B| = 2^5 = 32.$$

$$|A \cup B| = 256 + 128 - 32 = 352.$$

Answer: 352 strings meet the condition.

—

3. Stretch: A database lists 420 students in biology, 380 in chemistry, and 160 in both. How many distinct students are in at least one science course?

Solution:

$$420 + 380 - 160 = 640.$$

Answer: 640 unique students.

—

8.5 Python Demonstration

Listing 8.1: Demonstrating the Subtraction Rule in Python

```
1 """
2 COMSC_2043_-_Subtraction_Rule_Demonstrations
3 Author:_Jeremy_Evert
4 """
5
6 def subtraction_rule(n1, n2, overlap):
7     """Return_n1+_n2-_overlap."""
8     return n1 + n2 - overlap
9
10 # Example 1: 8-bit strings
11 A, B, overlap = 2**7, 2**6, 2**5
12 print("8-bit_strings:", subtraction_rule(A, B, overlap))
13
14 # Example 2: Students in Math or CS
15 math, cs, both = 130, 85, 42
16 print("Unique_students:", subtraction_rule(math, cs, both))
```

—

Reflection

The subtraction rule guards against double-vision in counting. Whenever two sets overlap, the shared region must be removed once to restore accuracy.

From bit strings to course enrollments, the rule ensures each unique element is tallied only once—a quiet act of mathematical honesty that underpins databases, logic, and probability alike.

Chapter 9

Section 6.1.5 — The Division Rule

Why This Section Matters

Sometimes we intentionally overcount by labeling or rotating, and then need to *divide out* the symmetry. The **Division Rule** tells us when dividing by a constant number of equivalent descriptions yields the correct number of unique outcomes.

9.1 The Division Rule

Definition 9.1 (Division Rule). If each distinct outcome of a task is represented by exactly d different descriptions (that is, there is a d -to-1 mapping from descriptions to outcomes), and there are n total descriptions, then the number of distinct outcomes is:

$$\frac{n}{d}.$$

Set/function viewpoint. If A is the set of descriptions and B the set of outcomes, and $f: A \rightarrow B$ is onto with each $b \in B$ having exactly d preimages in A , then $|B| = |A|/d$.

Sanity checks and gotchas.

- The rule only works when *every* outcome has the *same* number d of representations.
- “Equivalent” must form an *equivalence relation* (reflexive, symmetric, transitive) so that A splits into equal-sized buckets.
- If different outcomes have different numbers of representations, you can’t use a single division—this is where tools like Burnside’s Lemma or the orbit–stabilizer theorem appear.

9.2 Worked Examples

Example 9.1 (Counting cows from legs (Rosen-flavored)). An automated system counts exactly 572 cow legs in a pasture. Assuming only cows are present and each cow has four legs, how many cows are in the pasture?

Solution. Descriptions A : individual *legs* ($n = 572$). Outcomes B : individual *cows*. Each cow contributes $d = 4$ leg-descriptions, so

$$|B| = \frac{572}{4} = 143.$$

Example 9.2 (Circular seatings up to rotation). How many distinct seatings of four distinct people around a circular table are there, if seatings that differ only by a rotation are considered the same?

Solution. Linear seatings: $4!$ descriptions. Rotating the circle doesn't change the pattern, and there are $d = 4$ rotations. Thus

$$\frac{4!}{4} = 3! = 6.$$

Example 9.3 (Unordered pairs from ordered pairs). How many undirected edges (unordered pairs) exist on n labeled vertices?

Solution. Ordered pairs of distinct vertices: $n(n - 1)$. Each undirected edge $\{u, v\}$ corresponds to exactly $d = 2$ ordered descriptions (u, v) and (v, u) , so

$$\binom{n}{2} = \frac{n(n - 1)}{2}.$$

Example 9.4 (Anagrams with repeated letters). How many distinct anagrams does the word BALLOON have?

Solution. Treat the seven positions as labeled: $7!$ descriptions. Swapping the two indistinguishable L 's and the two indistinguishable O 's doesn't change the outcome. Thus each unique arrangement has $d = 2! \times 2!$ equivalent descriptions:

$$\frac{7!}{2! 2!} = 1260.$$

—

9.3 Try It — Easier (solution on next page)

A warehouse robot counts 1,140 wheels on identical cargo carts. Each cart has exactly six wheels. Assuming no other wheeled objects are present, how many carts are there?

Solution. Each cart contributes $d = 6$ wheel-descriptions. By the Division Rule:

$$\frac{1,140}{6} = 190 \text{ carts.}$$

—

9.4 Challenge — Harder (solution on next page)

Seven distinct programmers sit around a round table. Two seatings are the same if one can be rotated into the other (neighbor relationships match). How many distinct seatings exist? Then generalize to n people.

Solution. Linear arrangements: $7!$. Each circular seating has exactly $d = 7$ rotations, so

$$\frac{7!}{7} = 6! = 720.$$

In general, for n people:

$$\frac{n!}{n} = (n-1)!.$$

Bonus: If reflections also count as identical (like a necklace flip), divide by 2 again for $n > 2$.

—

9.5 Python Demonstrations

Listing 9.1: Division Rule demos in Python

```

1 from math import factorial
2
3 def circular_seatings(n):
4     """Count seatings up to rotation:  $n! / n = (n-1)!$ """
5     return factorial(n) // n
6
7 def unordered_pairs(n):
8     """Count unordered pairs  $\{u, v\}$ :  $n(n-1)/2$ """
9     return n * (n - 1) // 2
10
11 print("Circular seatings for 4:", circular_seatings(4)) # 6
12 print("Circular seatings for 7:", circular_seatings(7)) # 720
13 for n in [5, 10, 100]:
14     print(f" $K_n$  has", unordered_pairs(n), "edges")

```

—

Checklist: When Can I Divide?

- I first counted *descriptions*.
- I identified an equivalence relation grouping them into outcomes.
- Each outcome has the same number d of descriptions.

- Therefore, $\text{outcomes} = \frac{\text{descriptions}}{d}$.

—

Reflection

The Product Rule multiplies possibilities. The Subtraction Rule removes overlap. The Division Rule removes symmetry. Together, they form a trinity of clarity for counting—and they echo everywhere in computer science: hash buckets, graph edges, canonical forms, and seating symmetries alike.

Chapter 10

Section 6.1.6 — Tree Diagrams

Why this section matters

When a counting problem starts to branch—when every choice opens new sub-choices—our mental arithmetic hits its limits. Tree diagrams visualize those branching worlds. Each path from the root to a leaf represents one distinct outcome, and the total number of leaves tells us how many possibilities exist. In computer science, these trees reappear as recursion traces, decision trees, and game trees—every “if \rightarrow else” path you’ve ever coded.

10.1 Tree Diagrams

Definition 10.1 (Tree Diagram). A **tree diagram** is a branching structure that represents sequential choices. Each vertex (node) shows a decision point, and each edge (branch) corresponds to an outcome of that decision. Every path from the root to a leaf represents one complete sequence of choices.

Tree diagrams make the invisible visible:

- Each branch multiplies choices (Product Rule).
- Each split visualizes conditional logic (“if/else” in code).
- The number of leaves equals the total number of possible outcomes.

10.2 Worked Examples

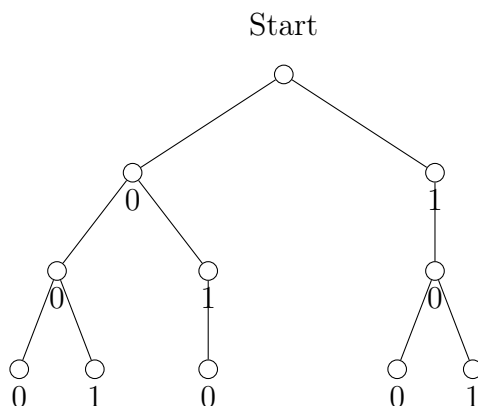
Example 10.1 (Bit strings without consecutive 1s (Rosen Example 22)). How many bit strings of length 4 contain no two consecutive 1s?

Solution. Draw a tree where each level adds one bit (0 or 1). When a 1 is chosen, the next level cannot start with another 1. This yields the valid strings:

0000, 0001, 0010, 0100, 0101, 1000, 1001, 1010.

Hence, 8 valid bit strings in total.

Visual Tree Representation:



Each branch shows a legal continuation. Every leaf (bottom node) corresponds to a valid 4-bit string without consecutive 1s.

Example 10.2 (Playoff possibilities (Rosen Example 23)). A playoff between two teams consists of at most five games. The first team to win three games wins the playoff. How many different ways can the playoff unfold?

Solution. Each game outcome is either W (win) or L (loss). The series stops as soon as one team reaches 3 wins. Enumerating all branches that terminate at “3 wins” yields 20 distinct series paths. Thus, there are 20 ways the playoff can occur.

Example 10.3 (T-shirt varieties (Rosen Example 24)). Suppose a souvenir shop sells “I ♣ New Jersey” T-shirts in five sizes (S, M, L, XL, XXL) and three colors (white, red, black). How many distinct shirts must be stocked?

Solution. Each shirt is determined by one choice of size and one choice of color. The tree’s first branch: five sizes. Each size branch splits into three color options. By the Product Rule:

$$5 \times 3 = 15$$

distinct shirts must be stocked.

10.3 Try It — Easier (solution on the next page)

A restaurant offers 3 appetizers, 4 main courses, and 2 desserts. How many full three-course meals can be made if you choose one of each?

Solution. Each appetizer branch splits into 4 main courses, each of which splits into 2 desserts:

$$3 \times 4 \times 2 = 24.$$

Hence, 24 possible full meals.

10.4 Challenge — Harder (solution on the next page)

A security code is three characters long. Each character is a letter (A–Z) or a digit (0–9). However, a code may not contain two consecutive digits. How many valid codes exist?

Solution. We use a branching tree or recursive reasoning.

Let L_n = codes of length n ending with a letter, and D_n = codes of length n ending with a digit.

Recurrence:

$$L_n = 26(L_{n-1} + D_{n-1}), \quad D_n = 10L_{n-1}.$$

Base case: $L_1 = 26$, $D_1 = 10$.

Compute:

$$\begin{aligned} L_2 &= 26(26 + 10) = 936, & D_2 &= 10(26) = 260, \\ L_3 &= 26(936 + 260) = 31,096, & D_3 &= 10(936) = 9,360. \end{aligned}$$

Total = $L_3 + D_3 = 40,456$ valid codes.

10.5 Python Demonstration

Listing 10.1: Tree diagram counting via recursion

```

1 def codes(n):
2     """Count length-n codes with no consecutive digits."""
3     if n == 1:
4         return 26, 10 # (end-with-letter, end-with-digit)
5     L_prev, D_prev = codes(n-1)
6     L = 26 * (L_prev + D_prev)
7     D = 10 * L_prev
8     return L, D
9
10 L3, D3 = codes(3)
11 print("Valid 3-char codes:", L3 + D3)
```

Reflection

Tree diagrams transform abstract multiplication and subtraction rules into visual reasoning. They help debug logic—each branch exposes a case that might otherwise hide in algebra. For programmers, they mirror recursive call trees and state machines. When you can see the structure, you can count the structure.

Counting Toolkit Summary

- **Sum Rule:** If choices are disjoint, $\text{total} = n_1 + n_2$.
- **Product Rule:** For sequential decisions, $\text{total} = n_1 \times n_2$.
- **Subtraction Rule:** If overlap exists, $\text{total} = n_1 + n_2 - n_{1,2}$.
- **Complement Rule:** Count “total – forbidden.”
- **Permutation:** Ordered selection of r from n : $P(n, r) = \frac{n!}{(n-r)!}$.
- **Combination:** Unordered selection of r from n : $C(n, r) = \frac{n!}{r!(n-r)!}$.
- **Tree Diagrams:** Visualize sequential branching; total outcomes = number of leaves.

Counting is reasoning. Each rule is a way to avoid double-counting and to see structure in the infinite forest of possibilities. ♣

Chapter A

Episode 1 — Counting the Possibilities (Podcast Transcript)

[Soft Intro Music Fades In ★]

Amina (Philosopher): Hello, friends — and welcome to today’s *Discrete Structures Podcast*! I’m Dr. Amina Reyes, your host and fellow explorer of mathematical mysteries. Today, we’re diving into one of my favorite topics: **counting**. Not sheep, not calories — but possibilities.

Joining me in the studio are three incredible minds from different corners of our learning universe.

Amina: We’ve got Lin Tran, our resident coder and problem-solver from California — always ready to turn theory into Python.

Lin (Coder): Hey, everyone! I brought code, coffee, and way too many parentheses.

Amina: Next, the dreamer of our group — Zahra Patel, who sees logic like poetry.

Zahra (Dreamer): Hi, friends. I see counting as choreography — numbers dancing in patterns we can learn to follow.

Amina: And of course, we couldn’t do this without our brave skeptic, Jake Miller — a first-year student who keeps us honest.

Jake (Skeptic): “Brave” or “confused,” depending on the day — but I’m here and caffeinated.

—

[Music Fades Out ■]

Amina: So, team — let's start with a simple question. Why do we count?

Jake: Because math professors make us?

Lin: (laughing) That's fair. But also — because we want to *predict*. Counting is how we map what could happen before it happens.

Zahra: It's like peeking into parallel universes — each choice we make branches into another possible outcome.

Amina: Beautifully said. Mathematicians call that the **Rule of Product** — each choice multiplies the number of possibilities. So if you pick an outfit: 3 shirts and 2 pairs of pants, that's $3 \times 2 = 6$ combinations.

Jake: Okay, but what if I wear both pairs of pants at once? Asking for a friend.

Lin: Then you've entered a new dimension of fashion — and combinatorics.

—

[Light Background Music ★]

Amina: Now, sometimes we count by adding instead of multiplying — that's the **Rule of Sum**. If a student can take either a math elective or a computer science elective — two separate paths — we add their options together.

Zahra: It reminds me of choosing between two adventures. Either one is exciting — you just can't live both at once.

Jake: So multiplying is “this *and* that,” and adding is “this *or* that”?

Lin: Exactly! You're getting it. And when you mix those ideas — “this and that or maybe that too” — you get trees, networks, recursion... basically, everything you'll ever debug at 2 AM.

Amina: (laughing) Spoken like a true coder.

—

[Pause — Music Fades Out ✱]

Amina: But what happens when our counts overlap — when some possibilities get counted twice?

Jake: Oh, like when you RSVP for the same party on two apps?

Amina: Exactly. That's when we use the **Subtraction Rule** to remove duplicates.

Zahra: A little mathematical honesty — no double-dipping in the data.

Lin: I like to think of it as de-bugging reality.

Amina: And from there, we climb higher — to permutations, combinations, and probability. Counting is the seed of it all.

—

[Reflective Outro Music ★]

Amina: Before we wrap up, I'd love a closing thought from each of you.

Lin: Counting shows that every little choice — every “if” in a line of code — matters.

Zahra: To me, counting is comfort. It tells us that even in complexity, patterns exist — and they're beautiful.

Jake: I'll admit it — counting's cooler than I thought. Especially when it keeps me from writing 40 test cases by hand.

Amina: (laughing) Progress! Well, thank you all — Lin, Zahra, and Jake — for making math a little more human today.

And to our listeners: whether you're a dreamer, a coder, or a skeptic yourself — remember, counting isn't about numbers. It's about noticing possibility.

Amina (softly): Until next time, keep wondering, keep counting, and keep being kind to your brain.

[Outro Music Swells and Fades ★]

—

Production Notes (for TTS Integration)

- Script voices correspond to those in `data/personas.json`.
- Insert small pauses between section breaks for natural pacing.
- Each [Music ...] line can later map to intro/outro sound cues.