

# ALGORITHM COMPLEXITIES

Understanding How Things Scale in Everyday Life

CS 101 - Fall 2025

**Welcome to Algorithm Complexities!**



## Today's Mission

Learn the **5 most important complexity levels** that describe how things scale in real life!

### 💡 Tip

What we'll discover:

- $O(1)$  - The Magic Trick Level
- $O(\log n)$  - The Smart Detective Level
- $O(n)$  - The One-by-One Level
- $O(n^2)$  - The Handshake Problem Level
- $O(2^n)$  - The Explosion Level

Ready to become complexity detectives? Let's go!

## Complexity is All About How Things Scale

### Note

**Complexity** = How much **more work** do you need when you have **more stuff** to deal with?

**Real-Life Examples:**

- **Making dinner for friends:** 2 friends vs 20 friends - how much more work?
- **Finding a book:** In a small pile vs a huge library - how much longer?
- **Gift wrapping:** 5 gifts vs 50 gifts - how much more time?
- **Meeting everyone at a party:** 10 people vs 100 people - how many more handshakes?

## The Big Question


**! When you double the amount of “stuff,” what happens to the amount of work?**

- Does it stay the same?
- Double too?
- Get **way** worse?
- Or explode completely?

**That's what complexity tells us!**

Let's explore each complexity level!

### **O(1) - The Magic Trick Level**

 **“No Matter How Much, It Takes the Same Time!”**

**O(1) means:** Whether you have 1 thing or 1 million things, the task takes exactly the same amount of time!

**i** **Everyday  $O(1)$  Examples:**  
Using a key to open your door - Same one turn always!  
Turning on a light switch - Same flip always!  
Checking the time on your phone - Always instant!  
Using your debit card - Same swipe time always!

## Why $O(1)$ is Amazing

### **The Holy Grail of Algorithms!**

It's like magic - the amount of work **never changes**  
Perfect performance - always fast, always reliable  
Every programmer dreams of  $O(1)$  solutions!

### **i** **Real-World $O(1)$ Examples:**

- Your phone's "Recent Calls" list
- Looking up a contact by name
- Checking account balance
- Skipping to specific song

## $O(\log n)$ - The Smart Detective Level

### **"Cut the Problem in Half, Over and Over!"**

**$O(\log n)$  means:** Each step eliminates half of what's left to search. Super efficient even with huge amounts!

### **i** **Everyday $O(\log n)$ Examples:**

**Guessing a number 1-1000** - Cut problem in half each time, found in ~10 questions max!  
**Finding word in dictionary** - Open to middle, go left or right, found in seconds!  
**20 Questions game** - Each question eliminates half the possibilities  
**Phone contact search** - Type "J" → cuts to J names, type "Jo" → even fewer options

## Why $O(\log n)$ is Amazing

### 💡 Incredible Scaling Performance!

**Amazing scaling:** \* 1,000 items  $\rightarrow$  ~10 steps \* 1,000,000 items  $\rightarrow$  ~20 steps

\* 1,000,000,000 items  $\rightarrow$  ~30 steps

**Smart strategy beats brute force**

### i Used everywhere:

- Google searches
- GPS route finding
- Phone contact search

## But What's the Catch?



### ⚠ The catch: You need things **organized first!**

## $O(n)$ - The One-by-One Level

**i** “Check Every Single Thing, One by One”

**$O(n)$  means:** Double the stuff = Double the work. Fair and predictable!

**i** Everyday  $O(n)$  Examples:

**Reading every page in a book** - 100 pages = 100 page flips, 200 pages = 200 page flips

**Counting items in shopping cart** - Must touch each item once, 10 items = 10 counts

**Listening to playlist** - 50 songs =  $50 \times$  the time

**Grading test stack** - 30 tests =  $30 \times$  the work

## Why $O(n)$ is Pretty Good

**i** Predictable and Fair!

**Predictable and fair** - work scales linearly

**Often the best you can do** when you need to check everything

**Reasonable for most tasks:** \* Finding highest grade \* Adding up expenses \* Reading all text messages

**⚠ When it gets slow:**

Really large amounts of data - but still very manageable for normal use!

## $O(n^2)$ - The Handshake Problem Level

**⚠ “Everyone Must Meet Everyone Else!”**

**$O(n^2)$  means:** When you double the people, you get **four times** the work! This gets crazy fast.

**i Everyday  $O(n^2)$  Examples:**

**Party introductions** - 4 people = 6 handshakes, 8 people = 28 handshakes, 16 people = 120 handshakes!

**Sports tournament** - Everyone plays everyone, gets expensive fast!

**Group photo arrangements** - Every person next to every other, gets overwhelming quickly!

**Comparing all student tests** - Looking for identical answers, 30 students = 435 comparisons!

## Why $O(n^2)$ Gets Scary

**⚠ Explosive Growth!**

**Explosive growth:** \* 10 things  $\rightarrow$  100 operations \* 100 things  $\rightarrow$  10,000 operations \* 1,000 things  $\rightarrow$  1,000,000 operations!

**The danger zone** - where apps become unusably slow

**i Common culprits:** \* Comparing every item to every other \* Nested loops in programming \* Poor algorithm choices

**! When to worry:** Anything over ~1,000 items gets really slow!

## $O(2)$ - The Explosion Level

**! “Every Choice Doubles Your Problems!”**

**$O(2)$  means:** Add just one more thing, and you **double** all the work! This explodes instantly.

**i Everyday  $O(2)$  Examples:**  
**Family tree exploration** - 2 parents  $\rightarrow$  4 grandparents  $\rightarrow$  8 great-grandparents  $\rightarrow$  16 great-great-grandparents  
**Password cracking** - Each digit doubles possibilities, 10-digit PIN = 1+ billion combos!  
**Gift wrapping combinations** - Each gift: wrapped or not, 20 gifts = 1+ million combinations!

## Why $O(2)$ is Terrifying

### ! Grows Impossibly Fast!

**Grows impossibly fast:** \* 10 things  $\rightarrow$  1,024 operations \* 20 things  $\rightarrow$  1,048,576 operations

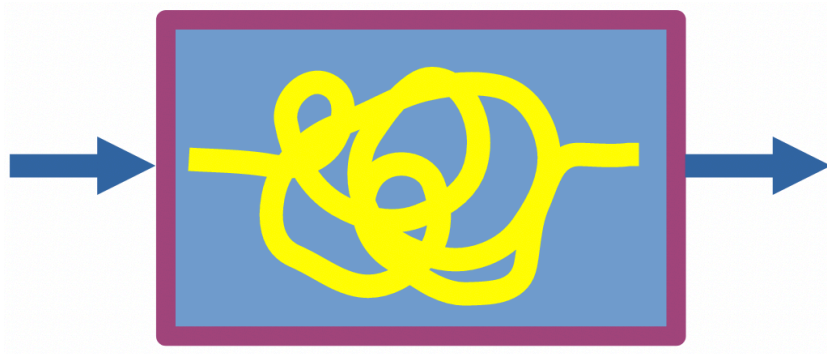
\* 30 things  $\rightarrow$  1,073,741,824 operations!

Usually unusable for anything but *tiny* problems

**i Real-world impact:** \* Why cryptography works (good!) \* Why some problems are “impossible” (bad!)

**! Bottom line:** Avoid at all costs unless you have  $< 20$  items!

## The Complexity Race!



### 💡 How They Compare With 1,000 Items

Let's see what happens when we have 1,000 things to process:

## The Complexity Race Table

Complexity	Name	Steps Needed	Real-World Feeling
$O(1)$	Magic Trick	<b>1 step</b>	Instant!
$O(\log n)$	Smart Detective	<b>~10 steps</b>	Super fast!
$O(n)$	One-by-One	<b>1,000 steps</b>	Takes a moment
$O(n^2)$	Handshake Problem	<b>1,000,000 steps</b>	Ugh, so slow...
$O(2^n)$	Explosion	<b><math>2^1</math> steps</b>	Heat death of universe

## The Big Takeaway

❗ **Small differences in complexity = HUGE differences in real-world performance!**

This is why choosing the right approach matters so much in programming!

Ready for your challenge? Let's become complexity detectives!

## Your Turn: Complexity Detectives!

**i** **Now You're Ready for the Challenge!**

You've learned the 5 complexity levels. Time to become **complexity detectives** and find examples from your own life!



## The “Build a Better Algorithm” Challenge

### Tip

#### Your Mission:

1. **Brainstorm** real-life situations that match each complexity level
2. **Work in teams** to find creative examples
3. **Think about** when you’d choose one approach over another
4. **Share** your discoveries with the class!

## Remember the Levels!

### The 5 Complexity Levels:

- $O(1)$  - Magic Trick (always same time)
- $O(\log n)$  - Smart Detective (cut in half)
- $O(n)$  - One-by-One (check everything)
- $O(n^2)$  - Handshake Problem (everyone meets)
- $O(2^n)$  - Explosion (choices double work)

## Questions for Detective Work!

### Questions to Ask Yourself:

- What happens when I double the input?
- Do I compare everything to everything?
- Can I organize data for faster searching?
- Am I exploring all combinations?

**Pro Tip:** Look for efficient patterns in your daily life!

Let’s see what amazing complexity examples you can find!