

Asymptotic Notation: Conceptual

What is Asymptotic Notation?

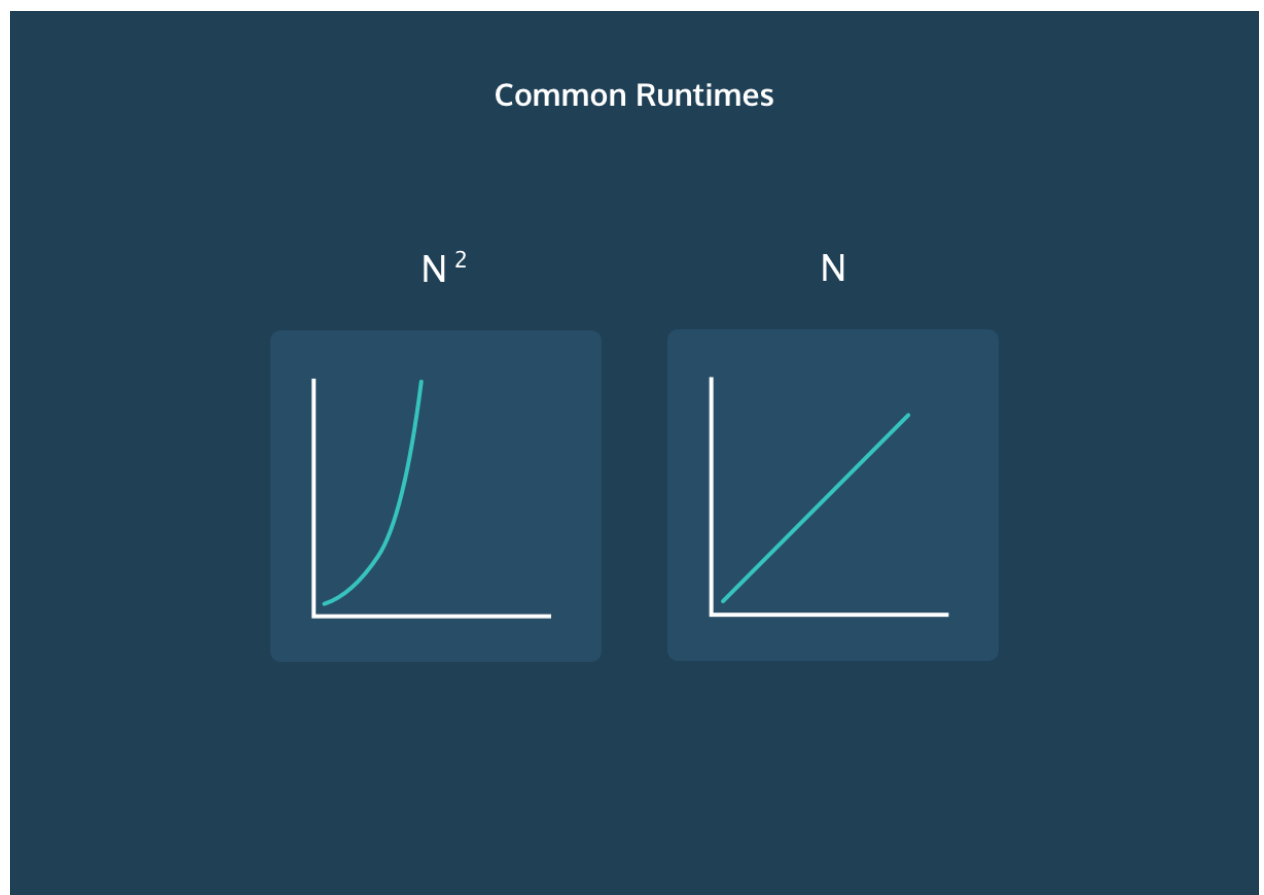
6 min

Cheetahs. Ferraris. Life. All are fast, but how do you know which one is the fastest? You can measure a cheetah's and a Ferrari's speed with a speedometer. You can measure life with years and months.

But what about computer programs? In fact, you *can* time a computer program, but different computers run at different speeds. For example, a program that takes 12 nanoseconds on one computer could take 45 milliseconds on another. Therefore, we need a more general way to gauge a program's runtime. We do this with *Asymptotic Notation*.

Instead of timing a program, through asymptotic notation, we can calculate a program's runtime by looking at how many instructions the computer has to perform based on the size of the program's input: N .

For instance, a program that has input of size N may tell the computer to run $5N^2+3N+2$ instructions. (We will get into how we get this kind of expression in future exercises.) Nevertheless, this is still a fairly messy and large expression. For asymptotic notation, we drop all of our constants (the numbers) because as N becomes extremely large, the constants will make minute differences. After changing our constants, we have N^2+N . If we take each of these terms in the expression and graph them, we see that the N^2 term grows faster than the N term.



For example, when N is 1000:

- the N^2 term is 1,000,000
- the N term is 1,000

As you can see, the N^2 term is much more significant than the N term. When N is larger than 1000, the difference becomes even more significant. Because the difference is so enormous, we don't even need to consider the N term when calculating the runtime. Thus, for this program, we would describe the runtime in terms of N^2 . There are three different ways we could describe the runtime of this program: big Theta or $\Theta(N^2)$, big O or $O(N^2)$, big Omega or $\Omega(N^2)$. The difference between the three and when to use which one will be detailed in the next exercises.

You may see the term **execution count** used in evaluating algorithms. Execution count is more precise than Big O notation. The following method, `addUpTo()`, depending on how we count the number of operations, can be as low as $2N$ or as high as $5N + 2$

```
public class Main() {  
    void int addUpTo(int n) {  
        int total = 0;  
        for (int i = 1; i <= n; i++) {  
            total += i;  
        }  
        return total;  
    }  
}
```

Determining execution count can increase in difficulty as our algorithms become even more sophisticated!

But regardless of the execution count, the number of operations grows roughly proportionally with n . If n doubles, the number of operations will also roughly double.

Big O Notation is a way to formalize fuzzy counting. It allows us to talk formally about how the runtime of an algorithm grows as the inputs grow. As we will see, Big O doesn't focus on the details, only the trends

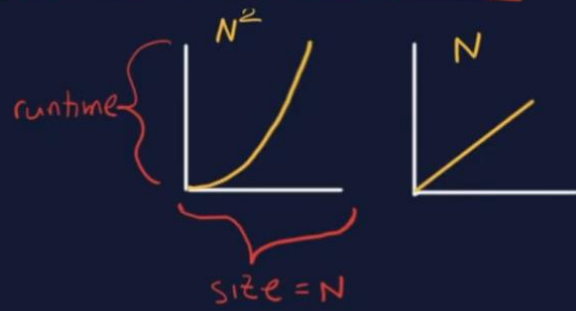
Instructions

Play the video to learn about asymptotic notation in a more visual setting.

ASYMPTOTIC NOTATION

function(input):

size = N



$$N^2 + N$$

$$5N^2 + 2N + 4$$

$$\Theta(N^2)$$

$$O(N^2)$$

$$\Omega(N^2)$$