CS 61B Spring 2018

More Asymptotic Analysis

Discussion 8: March 6, 2018

Here is a review of some formulas that you will find useful when doing asymptotic analysis.

- $\sum_{i=1}^{N} i = 1 + 2 + 3 + 4 + \dots + N = \frac{N(N+1)}{2} = \frac{N^2 + N}{2}$
- $\sum_{i=0}^{N-1} 2^i = 1 + 2 + 4 + 8 + \dots + 2^{N-1} = 2 \cdot 2^{N-1} 1 = \mathbf{2^N} \mathbf{1}$

The goal of this discussion is to help us analyze in trees, but I didn't get it...

Intuition

}

For the following recursive functions, give the worst case and best case running time in the appropriate $O(\cdot)$, $\Omega(\cdot)$, or $\Theta(\cdot)$ notation.

 $\boxed{1.1}$ Give the running time in terms of N.

next question

2

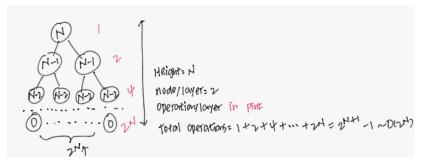
 $\fbox{1.2}$ Give the running time for andwelcome(arr, 0, N) where N is the length of the input array arr.

```
public static void andwelcome(int[] arr, int low, int high) {
        System.out.print("[ ");
                                                                      Big Theta(N or logN)
2
        for (int i = low; i < high; i += 1) {</pre>
3
             System.out.print("loyal ");
        }
        System.out.println("]");
        if (high - low > 0) {
7
            double coin = Math.random();
8
            if (coin > 0.5) {
                andwelcome(arr, low, low + (high - low) / 2);
10
                                                                      two lines here!
            } else {
11
                andwelcome(arr, low, low + (high - low) / 2);
12
                andwelcome(arr, low + (high - low) / 2, high);
13
            }
14
        }
15
    }
16
```

 $\boxed{1.3}$ Give the running time in terms of N.

```
public int tothe(int N) {
    if (N <= 1) {
        return N;
    }
    return tothe(N - 1) + tothe(N - 1);
}</pre>
```

Big Theta(2^N)



Give the running time in terms of N.

```
Worst: Big Theta(N*N!)
   public static void spacejam(int N) {
       if (N <= 1) {
2
            return;
3
       }
       for (int i = 0; i < N; i += 1) {
5
            spacejam(N - 1);
       }
   }
                                                              Height: N
                                                              node/layer: in pink
                                                               operation/layer in pink
                                                HIT
```

Hey you watchu gon do

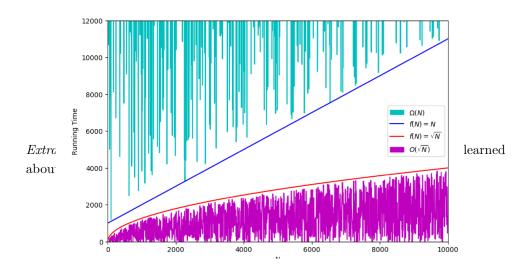
- 2.1 For each example below, there are two algorithms solving the same problem. Given the asymptotic runtimes for each, is one of the algorithms **guaranteed** to be faster? If so, which? And if neither is always faster, explain why.
 - (a) Algorithm 1: $\Theta(N)$, Algorithm 2: $\Theta(N^2)$ For large input, Algorithm 2 is faster
 - (b) Algorithm 1: $\Omega(N)$, Algorithm 2: $\Omega(N^2)$ No quarantee
 - (c) Algorithm 1: O(N), Algorithm 2: $O(N^2)$ No guarantee
 - (d) Algorithm 1: $\Theta(N^2)$, Algorithm 2: $O(\log N)$ For large input, Algorithm 2 is faster
 - (e) Algorithm 1: $O(N \log N)$, Algorithm 2: $\Omega(N \log N)$ For large input, Algorithm 1 is faster No, they can both be Big Theta(N log N)

Would your answers above change if we did not assume that N was very large (for example, if there was a maximum value for N, or if N was constant)?

Sure

Asymptotic Notation

3.1 Draw the running time graph of an algorithm that is $O(\sqrt{N})$ in the best case and $\Omega(N)$ in the worst case. Assume that the algorithm is also trivially $\Omega(1)$ in the best case and $O(\infty)$ in the worst case.



3.2 Are the statements in the right column true or false? If false, correct the asymptotic notation $(\Omega(\cdot), \Theta(\cdot), O(\cdot))$. Be sure to give the tightest bound. $\Omega(\cdot)$ is the opposite of $O(\cdot)$, i.e. $f(n) \in \Omega(g(n)) \iff g(n) \in O(f(n))$.

Fall 2015 Extra

- 4.1 If you have time, try to answer this challenge question. For each answer true or false. If true, explain why and if false provide a counterexample.
 - (a) If $f(n) \in O(n^2)$ and $g(n) \in O(n)$ are positive-valued functions (that is for all n, f(n), g(n) > 0), then $\frac{f(n)}{g(n)} \in O(n)$.

False; like f(n) is Big Theta (n^2) , g(n) is Big Theta(1), then f(n)/g(n) is actually Big Theta(n^2)

(b) If $f(n) \in \Theta(n^2)$ and $g(n) \in \Theta(n)$ are positive-valued functions, then $\frac{f(n)}{g(n)} \in \Theta(n)$.

True

Theorem 1. If $f(n) \in \Theta(n^2)$ and $g(n) \in \Theta(n)$ are positive-valued functions, then $\frac{f(n)}{g(n)} \in \Theta(n)$.

Proof. Given that $f \in \Theta(n^2)$ is positive, by definition there exists $k_0, k'_0 > 0$ such that for all n > N, the following holds.

$$k_0 n^2 \le f(n) \le k_0' n^2$$

Similarly, $g \in \Theta(n)$ implies there exists $k_1, k'_1 > 0$ such that

$$k_1 n \le g(n) \le k_1' n$$

Now consider $\frac{f(n)}{g(n)}$.

$$\frac{f(n)}{g(n)} \leq \frac{k_0' n^2}{k_1 n} = \frac{k_0' n}{k_1} \in O(n) \qquad \qquad \frac{f(n)}{g(n)} \geq \frac{k_0 n^2}{k_1' n} = \frac{k_0 n}{k_1'} \in \Omega(n)$$

As $\frac{f(n)}{g(n)}$ is in O(n) and $\Omega(n)$ then it is in $\Theta(n)$.