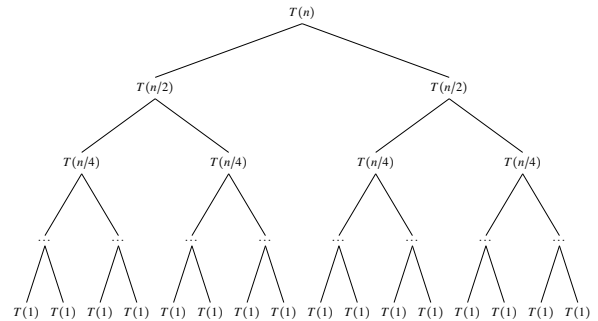


CSE 332

Data Structures and Parallelism

Algorithm Analysis 2



Outline

- 1 Summations
- 2 Warm-Ups
- 3 Analyzing Recursive Code
- 4 Generating and Solving Recurrences

Some Common Series

1

- Gauss' Sum: $\sum_{i=0}^n i = \frac{n(n+1)}{2}$
- Infinite Geometric Series: $\sum_{i=0}^{\infty} x^i = \frac{1}{1-x}$, when $|x| < 1$.
- Finite Geometric Series: $\sum_{i=0}^n x^i = \frac{1-x^{n+1}}{1-x}$, when $x \neq 1$.

Warm-Up #1: append

2

Let x and L be LinkedList Nodes.

Analyzing append

```

1 append(x, L) {
2   Node curr = L;
3   while (curr != null && curr.next != null) {
4     curr = curr.next;
5   }
6   curr.next = x;
7 }
```

What is ...

- a lower bound on the time complexity of `append`?
 $\Omega(n)$, because we always **must** do n iterations of the loop.
- an upper bound on the time complexity of `append`?
 $\mathcal{O}(n)$, because we never do **more** than n iterations of the loop.

Since we can **upper** and **lower** bound the time complexity with the same complexity class, we can say `append` runs in $\Theta(n)$.

Merge

3

Pre-Condition: L_1 and L_2 are sorted.

Post-Condition: Return value is sorted.

Merge

```

1 merge(L1, L2) {
2   p1, p2 = 0;
3   While both lists have more elements:
4     Append the smaller element to L.
5     Increment p1 or p2, depending on which had the smaller element
6   Append any remaining elements from L1 or L2 to L
7   return L
8 }
```

What is the... (remember the lists are Nodes)

- best case # of comparisons of `merge`?
 $\Omega(1)$. Consider the input: $[0], [1, 2, 3, 4, 5, 6]$.
- worst case # of comparisons of `merge`?
 $\mathcal{O}(n)$. Consider the input: $[1, 3, 5], [2, 4, 6]$.
- worst case space usage of `merge`?
 $\mathcal{O}(n)$, because we allocate a constant amount of space per element.

Consider the following code:

Merge Sort

```

1 sort(L) {
2   if (L.size() < 2) {
3     return L;
4   }
5   else {
6     int mid = L.size() / 2;
7     return merge(
8       sort(L.subList(0, mid)),
9       sort(L.subList(mid, L.size()))
10    );
11  }
12 }
```

What is the worst case/best case # of comparisons of sort?

Yeah, yeah, it's $\mathcal{O}(n \lg n)$, but why?

What is a recurrence?

In CSE 311, you saw a bunch of questions like:

Induction Problem

Let $f_0 = 0, f_1 = 1, f_n = f_{n-1} + f_{n-2}$ for all $n \geq 2$. Prove $f_n < 2^n$ for all $n \in \mathbb{N}$.

(Remember the Fibonacci Numbers? You'd better bet they're going to show up in this course!)

That's a recurrence. That's it.

Definition (Recurrence)

A recurrence is a recursive definition of a function in terms of smaller values.

Let's start with trying to analyze this code:

LinkedList Reversal

```

1 reverse(L) {
2   if (L == null) { return null; }
3   else if (L.next == null) { return L; }
4   else {
5     Node front = L;
6     Node rest = L.next;
7     L.next = null;
8
9     Node restReversed = reverse(rest);
10    append(front, restReversed);
11  }
12 }
```

Notice that `append` is the same function from the beginning of lecture that had runtime $\mathcal{O}(n)$.

So, what is the time complexity of `reverse`?

We split the work into two pieces:

- Non-Recursive Work
- Recursive Work

LinkedList Reversal

```

1 reverse(L) {
2   if (L == null) { return null; } //O(1)
3   else if (L.next == null) { return L; } //O(1)
4   else {
5     Node front = L; //O(1)
6     Node rest = L.next; //O(1)
7     L.next = null; //O(1)
8
9     Node restReversed = reverse(rest);
10    append(front, restReversed); //O(n)
11  }
12 }
```

Non-Recursive Work: $\mathcal{O}(n)$, which means we can write it as $c_0 + c_1n$ for some constants c_0 and c_1 .

LinkedList Reversal

```

1 reverse(L) {
2   if (L == null) { return null; }
3   else if (L.next == null) { return L; }
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5     Node front = L;
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7     L.next = null;
8
9     Node restReversed = reverse(rest);
10    append(front, restReversed);
11  }
12 }
```

Non-Recursive Work: $\mathcal{O}(n)$, which means we can write it as $c_0 + c_1n$ for some constants c_0 and c_1 .

Recursive Work: The work it takes to do `reverse` on a list one smaller. Putting these together almost gives us the recurrence:

$$T(n) = c_0 + c_1n + T(n-1)$$

We're missing the base case!

LinkedList Reversal

```

1 reverse(L) {
2   if (L == null) { return null; }
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5     Node front = L;
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9     Node restReversed = reverse(rest);
10    append(front, restReversed);
11  }
12 }
```

$$T(n) = \begin{cases} d_0 & \text{if } n = 0 \\ d_0 & \text{if } n = 1 \\ c_0 + c_1n + T(n-1) & \text{otherwise} \end{cases}$$

Now, we need to **solve** the recurrence.

$$T(n) = \begin{cases} d_0 & \text{if } n = 0 \\ d_1 & \text{if } n = 1 \\ c_0 + c_1 n + T(n-1) & \text{otherwise} \end{cases}$$

$$\begin{aligned} T(n) &= (c_0 + c_1 n) + T(n-1) \\ &= (c_0 + c_1 n) + (c_0 + c_1(n-1)) + T(n-2) \\ &= (c_0 + c_1 n) + (c_0 + c_1(n-1)) + (c_0 + c_1(n-2)) + \dots + (c_0 + c_1(2)) + d_0 + d_0 \\ &= \sum_{i=0}^{n-2} (c_0 + c_1(n-i)) + 2d_0 \\ &= \sum_{i=0}^{n-2} c_0 + \sum_{i=0}^{n-2} c_1(n-i) + 2d_0 \\ &= (n-1)c_0 + c_1 \sum_{i=1}^{n-1} i + 2d_0 \\ &= (n-1)c_0 + c_1 \left(\frac{(n-1)n}{2} \right) + 2d_0 \\ &= \mathcal{O}(n^2) \end{aligned}$$

A recurrence where we solve some constant piece of the problem (e.g. "-1", "-2", etc.) is called a **Linear Recurrence**.

We solve these like we did above by **Unrolling the Recurrence**.

This is a fancy way of saying "plug the definition into itself until a pattern emerges".

Today's Takeaways!



- Understand that Big-Oh is just an "upper bound" and Big-Omega is just a "lower bound"
- Know how to make a recurrence from a recursive program
- Understand what a linear recurrence is
- Be able to find a closed form linear recurrences
- Know the common summations