

Lecture 15: Asymptotics 2

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Loops

Loops Example 1: Based on Exact Count

- Find order of growth of worst case runtime:

```
int N = A.length;
for (int i = 0; i < N; i += 1)
    for (int j = i + 1; j < N; j += 1)
        if (A[i] == A[j])
            return true;
return false;
```

- Worst case number of `==` operations:
 - Given by area of right triangle of side length $N-1$
 - Area is $\Theta(N^2)$

Loops Example 2

```
int N = A.length;
for (int i = 1; i < N; i = i * 2)
    for (int j = 0; j < i; j += 1)
        System.out.println("hello");
        int ZUG = 1 + 1;
return false;
```

- $C(N) = 1 + 2 + 4 + \dots + N = 2N - 1$, if N is a power of 2
- Number of prints lies between $0.5N$ and $2N$
- The runtime complexity is, in fact, $\Theta(N)$

No Magic Shortcut

Repeat After Me...

- There is no magic shortcut for these problems (well... usually)
 - Runtime analysis often requires careful thought
 - CS70 and CS170 will cover this in much more detail
 - This is not a math class, we expect you to know these:
 - $1 + 2 + 3 + \dots + N = N(N+1)/2 = \Theta(N^2)$
 - $1 + 2 + 4 + \dots + N = 2N - 1 = \Theta(N)$ (Where N is a power of 2)
 - Strategies:

- Find exact sum
- Write out examples
- Draw pictures
- Use geometric intuition

Recursion

Recursion (Intuitive)

```
public static int f3(int n) {  
    if (n <= 1)  
        return 1;  
    return f3(n-1) + f3(n-1);  
}
```

- Our time complexity is $\Theta(2^N)$
- Every time we increase N by 1, we double the work!

Recursion and Exact Counting

- Another approach: count number of calls to $f3$, given by $C(N)$
 - $C(1) = 1$
 - $C(2) = 1 + 2$
 - $C(N) = 1 + 2 + 4 + \dots + 2^{(N-1)} = 2(2^{(N-1)}) - 1 = 2^N - 1$
- Since work during each call is constant:
 - $R(N) = \Theta(2^N)$

Recursion and Recurrence Relations

- Count number of calls to $f3$, by a "recurrence relation"
 - $C(1) = 1$
 - $C(N) = 2C(N-1) + 1$
- More technical to solve. Won't do this in our course

Binary Search

Binary Search Intuitive

- Finding a key in a sorted array
 - Compare key against middle entry
 - Too small, go left
 - Too big, go right
 - Equal, found
- The runtime of binary search is $\Theta(\log_2(N))$
- Why? Problem size halves over and over until it gets down to 1

Binary Search Exact Count

- Find worst case runtime for binary search
 - What is $C(6)$, number of total calls for $N = 6$?
 - 3
 - Three total calls, where $N = 6$, $N = 3$, and $N = 1$
 - $C(N) = \text{floor}(\log_2(N)) + 1$
 - Since compares take constant time, $R(N) = \Theta(\text{floor}(\log_2(N)))$
 - This $f(N)$ is way too complicated. Let's simplify.
 - Three useful properties:
 - $\text{floor}(f(N)) = \Theta(f(N))$
 - The floor of f has the same order of growth as f
 - $\text{ceiling}(f(N)) = \Theta(f(N))$
 - The ceiling of f has the same order of growth as f
 - $\log_p(N) = \Theta(\log_q(N))$
 - logarithm base does not affect order of growth
 - Hence, $\text{floor}(\log_2(N)) = \Theta(\log N)$
 - Since each call takes constant time, $R(N) = \Theta(\log N)$

Binary Search (using Recurrence Relations)

- $C(0) = 0$
- $C(1) = 1$
- $C(N) = 1 + C((N-1)/2)$

Log Time is Really Terribly Fast

- In practice, logarithm time algorithms have almost constant runtimes
 - Even for incredibly huge datasets, practically equivalent to constant time

Merge Sort

Selection Sort: A Prelude to Mergesort

- Earlier in class we discussed a sort called selection sort:
 - Find the smallest unfixed item, move it to the front, and "fix" it
 - Sort the remaining unfixed items using selection sort
- Runtime of selection sort is $\Theta(N^2)$
 - Look at all N unfixed items to find smallest
 - The look at $N-1$ remaining unfixed
 - ...
 - Look at last two unfixed items
 - Done, $\text{sum } 2+3+4+5+\dots+N = \Theta(N^2)$
- Given that runtime is quadratic, for $N = 64$, we might say the runtime for selection sort is 2048 arbitrary units of time (AU)

The Merge Operation

- Given two sorted arrays, the merge operation combines them into a single sorted array by successively copying the smallest item from the two arrays into a target array

- What is the time complexity of the merge operation?
 - $\Theta(N)$
 - Why? Use array writes as cost model, merge does exactly N writes

Using Merge to Speed Up the Sorting Process

- Merging can give us an improvement over vanilla selection sort:
 - Selection sort the left half: $\Theta(N^2)$
 - Selection sort the right half: $\Theta(N^2)$
 - Merge the results: $\Theta(N)$
- $N = 64$: ~ 1088 AU
 - Merge: ~ 64 AU
 - Selection sort: $\sim 2 \cdot 512 = \sim 1024$ AU
- Still $\Theta(N^2)$, but faster since $N + 2 \cdot (N/2)^2 < N^2$
 - 1088 vs 2048 AU for $N=64$

Two Merge Layers

- Can do even better by adding a second layer of merges
 - Two layers of merges: ~ 640 AU

Example 5: Mergesort

- Mergesort does merges all the way down (no selection sort):
 - If array is of size 1, return
 - Mergesort the left half
 - Mergesort the right half
 - Merge the results
- Total runtime to merge all the way down: ~ 384 AU
 - Top layer: $\sim 64 = 64$ AU
 - Second layer: $\sim 32 \cdot 2 = 64$ AU
 - Third layer: $\sim 16 \cdot 4 = 64$ AU
 - Overall runtime in AU is $\sim 64k$, where k is the number of layers
 - $k = \log_2(64) = 6$, so ~ 384 total AU

Mergesort Order of Growth

- Mergesort has worst case runtime = $\Theta(N \log N)$
 - Every level takes $\sim N$ AU
 - Top level takes $\sim N$ AU
 - Top level takes $\sim N/2 + N/2 = \sim N$
 - etc. etc.
 - Thus, total runtime is $\sim Nk$, where k is the number of levels
 - Note that $k = \log_2(N)$
 - Overall runtime is $\Theta(N \log N)$

Linear vs. Linearithmic ($N \log N$) vs Quadratic

- $N \log N$ is basically as good as N , and is vastly better than N^2

- For $N = 1000000$, the $\log N$ is only 20

Summary

- Theoretical analysis of algorithm performance requires **careful thought**
 - There are **no magic shortcuts** for analyzing code
 - In our course, it's OK to do exact counting or intuitive analysis
 - Know how to sum $1 + 2 + 3 + \dots + N$ and $1 + 2 + 3 + \dots + N$
 - We won't be writing mathematical proofs in this class
 - Many runtime problems you'll do in this class resemble one of the five problems from today. See textbook, study guide, and discussion for more practice
 - This topic has one of the highest skill ceilings of all topics in the course
- Different solutions to the same problem may have different runtimes
 - N^2 vs. $N \log N$ is an enormous difference
 - Going from $N \log N$ to N is nice, but not a radical change