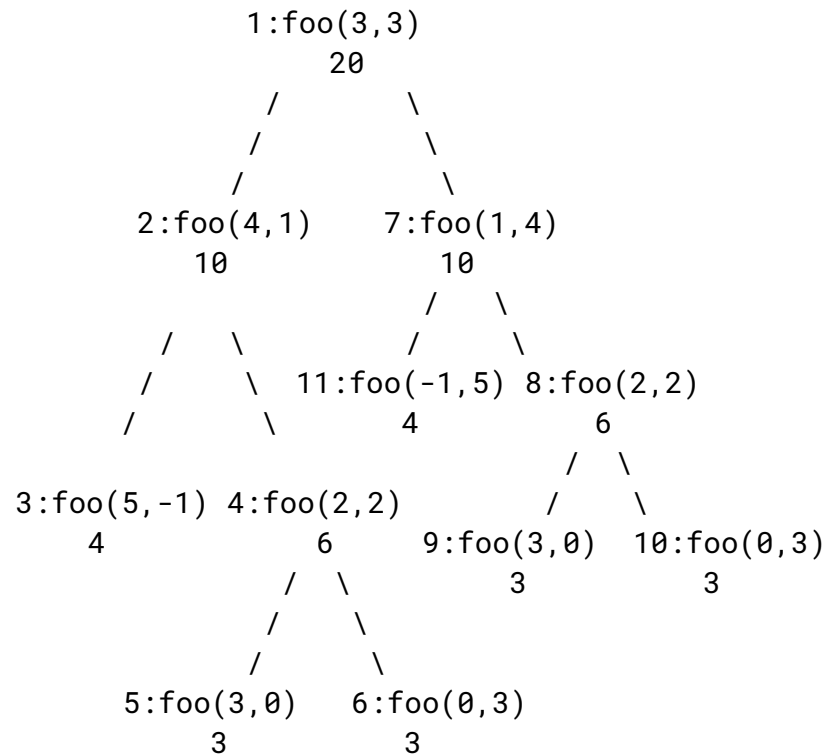


Problem Set 5, Part I

Problem 1: A method that makes multiple recursive calls

1-1)



1-2)

Call 3 (foo(5,-1)) returns 4
Call 5 (foo(3,0)) returns 3
Call 6 (foo(0,3)) returns 3
Call 4 (foo(2,2)) returns 6
Call 2 (foo(4,1)) returns 10
Call 9 (foo(3,0)) returns 3
Call 10 (foo(0,3)) returns 3
Call 8 (foo(2,2)) returns 6
Call 11 (foo(-1,5)) returns 4
Call 7 (foo(1,4)) returns 10
Call 1 (foo(3,3)) returns 20

Problem 2: Expressing Big-O

1. $a(n) = O(n)$
2. $b(n) = O(n^2)$
3. $c(n) = O(n^3)$
4. $d(n) = O(n \log(n))$
5. $e(n) = O(n^2)$
6. $f(n) = O(n^2)$
7. $g(n) = O(2^n)$

Problem 3: Computing Big-O

3-1) since the outer loop iterates n times and the middle loop iterates n times and the inner loop iterates n because it produces count if $k < j$. The big-O notation would be $n * n * n = O(n^3)$.

3-2) the outer loop iterates $\log(n)$ times, the middle loop iterates n times, and the $O(1000)$. The big-O notation is $\log(n) * n * 1000 = O(n \log(n))$

3-3) the outer loop iterates n times, the middle loop iterates $2n$ times, and the inner loop iterates $\log(n)$ times. The big-O notation would be $n * 2n * \log(n) = O(n^2 \log(n))$

3-4) the outer loop iterates $O(3)$ times, the middle loop iterates n times, and the inner loop iterates n times. The big-O notation would be $3 * n * n = O(n^2)$

3-5) The outer loop iterates n times and the inner loop n times as well. The big-O notation would be $n * n = O(n^2)$

Problem 4: Comparing two algorithms

4-1) The worst-case time efficiency for Algorithm A in terms of the length n of the array is $O(n^2)$ because it goes through the outer loop for the length of the array minus 1 times, which is n times, and it goes through the inner loop the length of the array times, which is also n . $n \cdot n = O(n^2)$

4-2) The worst-case time efficiency for Algorithm B is $O(n \log n)$ because mergesort has a worst-case efficiency of $O(n \log n)$ and the for loop iterates n times. $n + n \log n = O(n \log n)$

4-3)

```
public static int numDuplicatesC(int[] arr) {  
    int numDups = 0;  
    int copy = 0;  
    for (int i = 0; i < arr.length; i++) {  
        if (arr[i] == arr[copy]) {  
            numDups += 1;  
        }  
    }  
    return numDups;  
}
```

The worst case efficiency is $O(n)$ since the algorithm does not use any sorting and just runs the first and only loop which is $O(n)$.

Problem 5: Sum generator

5-1) $2n*(n(n+1))/2$

5-2) The efficiency of this algorithm is $O(n^3)$ because the simplification is n^3+n^2 and the slowest part of the algorithm is n^3 .

5-3)

```
public static String generateSums(int n) {
    String result = "";
    int sum = 0;
    String otherResult = "";

    for (int i = 1; i <= n; i++) {
        sum += i;

        if (n == 0) {
            return result;
        }

        if (i == 1) {
            otherResult += i;
            result += otherResult + "\n";
        } else if (i <= n) {
            otherResult += " + " + i;
            result += otherResult;
            if (i != n) {
                result += "\n";
            }
        }
    }

    result += " = " + sum;
    return result;
}
```

5-4) The time efficiency is $O(n)$ because the for loop iterates n times.

Problem 6: Basic Sorting Algorithms

6-1) The array looks like this after the second pass:

{3,4,18,24,33,40,8,10,12}

6-2) The array looks like this after the 4th iteration:

{4,10,18,24,33,40,8,3,12}

6-3) The array looks like this after the 3rd pass:

{4,10,18,8,3,12,24,33,40}

Problem 7: Comparing two algorithms

7-1) For algorithm A, the best case efficiency is $O(n)$ if there is only one value, but the average and worst case is $O(n)$ since it goes through all the elements n times.

7-2) For Algorithm B, The best case would be $O(n^2)$ since it would already be sorted but the worst and average case would be $O(n^2)$ since it goes through a nested for loop that has to iterate through.

7-3) Algorithm A would be more efficient since it has a faster run-time of $O(n)$.