Assignment 2: Dynamic Programming project

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October 30, 2012

1 Recursive function

```
struct maxS {
                    // Current sum
    int current;
                    // Overall max
    int max;
  struct maxS MaxSubarray(int array[], int size) {
    struct maxS ms;
                      // Base case
    if (size = 1) {
      ms.current = 0;
      ms.max = array[0];
    } else { // Recurse on array excluding last element.
12
      ms = MaxSubarray(array, size -1);
13
14
15
    // Find maximum.
    ms.current += array[size -1];
17
    ms.current = (ms.current > 0) ? ms.current : 0;
    ms.max = (ms.current > ms.max) ? ms.current : 0;
19
21
    return ms;
```

rec.cpp

Where maxS is a struct holding the running sum and the overall maximum sum and MaxSubarray is the recursive function. For an array A of size n, we find the maximum subarray with MaxSubarray(A, n).

2 Pseudocode

```
MaxSubarray(array, size):
 2
        current = 0;
        max = 0;
 3
        index = 0;
 5
 6
        while index < size:
 7
            current = current + array[index-1]
 8
            if current < 0:
9
                current = 0
10
            else if current > max:
11
                max = current
12
            index = index + 1
13
14
        return max
```

3 Running time

The code shows that we look at each element of the input array only once, so the algorithm's runtime should be $\Theta(n)$.

Figure 1 shows the execution time of this algorithm versus the size of the input array. The slope of the graph is 0.987956, which confirms the runtime of $\Theta(n)$.

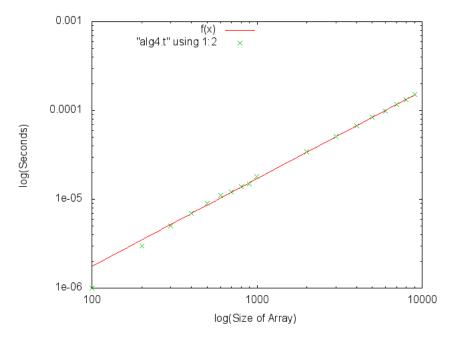


Figure 1: Algorithm 4 – execution time vs. input size

4 Theoretical correctness

Claim: MaxSubarray (A, n) will return the sum of the maximum subarray of an array A of size n.

Proof: As a base case, consider n = 0. Then max = current = 0, which is correct.

For the inductive hypothesis, assume that MaxSubarray(A, n) will correctly return the sum of the maximum subarray of an array A of size n. We must show that MaxSubarray(A, n+1) will do the same for an array A of size n+1.

We have two cases to consider:

Case 1: $A[n] \ge 0$. Since we are adding each element of the array to the current running sum on line 7, current will correctly increase and be correctly captured as the maximum sum on line 11.

Case 2: A[n] < -MaxSubarray(A, n). This will drive the running sum into the negative, so current will correctly "reset" to zero on line 9, effectively ignoring the sum up to element n. On the other hand, max retains its value, so the overall maximum sum is correctly preserved.

5 Implement

5.1 Algorithm 4

```
Enumeration
     Loop over each pair of indices i; j and compute the sum from k=i to j of a[k].
     Keep the best sum you have found so far.
  int MaxSubarray(int array[], int size){
    int current = 0;
    int max = 0;
10
    for (i=0; i < size; i++) {
12
       current += array[i];
13
       if (current \ll 0) {
14
15
         current = 0;
        else if (current > max) {
16
         \max = current;
18
19
20
21
    return max;
```

alg4.cpp

6 Test

Tests were run on the ms_test.txt file given for the last project and large arrays given by student IDs.

7 Compare

The two algorithms, one using divide-and-conquer and the other dynamic programming, were each tested using [TODO: How exactly was this tested?]. The results of the tests is shown in Figure 2.

The dynamic programming algorithm $(\Theta(n) \text{ time})$ is shown to be much more efficient than the divide-and-conquer algorithm $(O(n \log n) \text{ time})$ for a randomly sorted array. Additionally, since the DP algorithm does not have any recursive calls and needs to keep track of only two numbers (current and max), it is much less memory-intensive compared to the D&C algorithm.

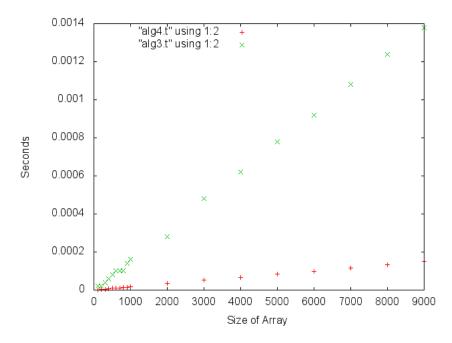


Figure 2: Comparison of execution times between the divide & conquer and dynamic programming algorithms.