

MTH-681 Analysis and Design of Algorithms
Assignment (3): Divide and Conquer
The Master Theorem

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3-5

Consider an $m \times n$ matrix of numbers, A , having the property that, for every $1 \leq i < j \leq m$ and $1 \leq k < l \leq n$, $A_{ik} + A_{jl} \leq A_{il} + A_{jk}$.

Let $lm(i)$ be the index of the column containing the leftmost minimum number of row i .

- a) Prove that, for every $1 \leq i < m$, $lm(i) \leq lm(i + 1)$.
- b) Describe a $O(m + n \log(m))$ divide-and-conquer algorithm that computes lm for all rows of A .

Solution:

a)

Proof. (By Contradiction)

Suppose for the sake of contradiction that $\forall i, j, k, l$:

$$1 \leq i < j \leq m \quad \wedge \quad 1 \leq k < l \leq n \implies A_{ik} + A_{jl} \leq A_{il} + A_{jk}$$

but $lm(i) > lm(i+1)$.

Let $j = i + 1$, $k = lm(i + 1) = lm(j)$, $l = lm(i)$

$\implies A_{il}$ is the minimum element of row i , and A_{jk} is the minimum element of row j .

$$\implies A_{il} \leq A_{ik} \quad \wedge \quad A_{jk} \leq A_{jl}$$

$$\implies A_{il} + A_{jk} \leq A_{ik} + A_{jl}$$

This is a contradiction to the assumption that $A_{ik} + A_{jl} \leq A_{il} + A_{jk}$.
Therefore, we must conclude that: $lm(i) \leq lm(i+1)$.

□

b)

```
// @code{matrix} is an mxn matrix with the property stated at the
// problem statement.
int[] leftMostMins(int[][] matrix) {
    int[] result = new int[matrix.length];
    return helper(matrix, 0, matrix.length - 1, 0, matrix[0].length - 1,
        result);
}

// Computes the left-most mins for @code{matrix} by recursing on the
// rows of the matrix and collecting the results into the
// @code{result} variable.
int[] helper(int[][] matrix, int lowRowIndex, int highRowIndex, int
    lowColumnIndex, int highColumnIndex, int[] result) {
    // Base case 1.
    if (lowRowIndex > highRowIndex) {
        return result;
    }

    // Base case 2.
    if (lowRowIndex == highRowIndex) {
        result[lowRowIndex] =
            findLeftMostMinIndexforSequence(matrix[lowRowIndex],
                lowColumnIndex, highColumnIndex);
        return result;
    }

    // Recursive step.
    // 1. Divide: find the index of the leftMostMin for midRow
    int midRowIndex = (lowRowIndex + highRowIndex) / 2;
    int midRowLeftMostMinIndex =
        findLeftMostMinIndexforSequence(matrix[midRowIndex],
            lowColumnIndex, highColumnIndex);

    // 2. Conquer: solve 2 smaller sub-problems
    helper(matrix, lowRowIndex, midRowIndex-1, lowColumnIndex,
        midRowLeftMostMinIndex, result);
    helper(matrix, midRowIndex+1, highRowIndex, midRowLeftMostMinIndex,
        highColumnIndex, result);

    // 3. Combine
    result[midRowIndex] = midRowLeftMostMinIndex;

    return result;
}
```

```
// Any  $O(\text{searchEndIndex} - \text{searchStartIndex})$  linear search algorithm.
int findLeftMostMinIndexforSequence(int[] sequence, int
    searchStartIndex, int searchEndIndex) {
    int result = searchEndIndex;
    int min = Integer.MAX_VALUE;
    for (int i = searchEndIndex; i >= searchStartIndex; i--) {
        if (sequence[i] <= min) {
            min = sequence[i];
            result = i;
        }
    }
    return result;
}
```

Correctness

Proof. (By Strong Induction)

Base case 1: $m = 0$

In this case the top call to helper is passed $lowRowIndex = 0$ and $highRowIndex = -1$, which hits base case 1 and correctly returns an empty array.

Base case 2: $m = 1$

In this case the top call to helper is passed $lowRowIndex = 0$ and $highRowIndex = 0$, which hits base case 2.

Inside base case 2, it correctly finds and sets the left-most min for that single row and returns the result.

Inductive step: $m > 1$

In the recursive step, the top call to helper is passed $lowRowIndex = 0$ and $highRowIndex > 0$, which hits the recursive step.

It then solves the problem at the middle row of the matrix, correctly sets the result for the middle row, and recurses on 2 smaller sub-problems, which by the induction hypothesis also set the result correctly for the remaining rows.

□

Average Case Time Complexity Analysis: $A_t(m, n)$

For simplicity, we do an average-case analysis assuming that *midRowLeftMostMinIndex* is equally likely to fall anywhere between the *lowColumnIndex* and *highColumnIndex*, and thus the left and right sub-problems are of nearly equal sizes.

The solution is then described by the following recurrence equation:

$$T(m, n) = 2T\left(\frac{m}{2}, \frac{n}{2}\right) + \Theta(n) + \Theta(1).$$

Notes:

1. The $\Theta(n)$ is contributed by the divide part of the algorithm.
2. The $\Theta(1)$ is included because we have to touch each row at least once.

The Master theorem does not apply straightforwardly to this formula, but it does apply to:

$$T_1(m, n) = 2T_1\left(\frac{m}{2}, \frac{n}{2}\right) + \Theta(n).$$

In this case: $a = 2, b = 2, f(n) = \Theta(n)$
 $\Rightarrow \lg_b a = \lg_2 2 = 1$
 $\Rightarrow n^{\lg_b a} = n^1 = n.$
 $\Rightarrow f(n) = \Theta(n^{\lg_b a})$
 \Rightarrow (Case 2)
 $\Rightarrow T_1(m, n) = \Theta(n^{\lg_b a} \lg m) = \Theta(n \lg m).$

Finally, we add the effect of the $\Theta(1)$ term in the original recurrence equation which gives an additional $\Theta(m)$ term in the solution:

$$T(m, n) = \Theta(n \lg m) + \Theta(m).$$