

In-class work: The Divide-and-conquer technique

The *maximum partial sum* (MPS) problem is defined as follows. Given an array A of n integers, find values of i and j with $0 \leq i \leq j < n$ such that

$$A[i] + A[i + 1] + \dots + A[j] = \sum_{k=i}^j A[k]$$

is maximized.

Example: For $A = [4, -5, 6, 7, 8, -10, 5]$, the solution to *MPS* is $i = 2$ and $j = 4$ ($6+7+8 = 21$).

(1) Consider the following array:

$$A = [13, -3, -25, 20, -3, -16, -23, 18, 20, -7, 12, -5, -22, 15, -4, 7]$$

Try to find MPS by hand. This will give an example of how MPS can include negative numbers.

(2) Describe a (straightforward) algorithm to find the MPS and analyze its running time.

As always, the question is: Can we do better? For e.g., can we solve MPS in $O(n \lg n)$ time?

As it turns out, a very neat $O(n \lg n)$ algorithm for MPS is possible via divide-and-conquer. We'll come up with it in a few steps.

(3) First, we consider the *left position ℓ maximal partial sum* problem ($LMPS_\ell$). Here the left index is given and the problem is to find the index j with $\ell \leq j \leq n$ such that

$$\sum_{k=\ell}^j A[k]$$

is maximized.

Example: For the array $[4,-5,6,7,8,-10,5]$ the solution to e.g. $LMPS_3$ is $j = 4$ ($7 + 8 = 15$). Describe $O(n)$ time algorithms for solving $LMPS_\ell$ for given ℓ .

(4) Similarly, the *right position r maximal partial sum* problem ($RMPS_r$), consists of finding value i with $1 \leq i \leq r$ such that

$$\sum_{k=i}^r A[k]$$

is maximized.

Example: For the array $[4,-5,6,7,8,-10,5]$ the solution to e.g. $RMPS_6$ is $i = 2$ ($5 - 10 + 8 + 7 + 6 = 16$).

Describe $O(n)$ time algorithms for solving $RMPS_r$ for given r .

(5) Using an $O(n)$ time algorithm for $LMPS_\ell$, describe a simple $O(n^2)$ algorithm for solving MPS.

(6) Using $O(n)$ time algorithms for $LMPS_\ell$ and $RMPS_r$, describe an $O(n \log n)$ divide-and-conquer algorithm for solving MPS .