

Lecture 2: Maximum Contiguous Subarray Problem

Overview

- Reference: Chapter 8 in *Programming Pearls*, (2nd ed) by Jon Bentley.
- Clean way to illustrate basic algorithm design
 - A $\Theta(n^3)$ **brute force** algorithm
 - A $\Theta(n^2)$ algorithm that **reuses data**.
 - A $\Theta(n \log n)$ **divide-and-conquer** algorithm
- *Cost* of algorithm will be number of primitive operations, e.g., comparisons and arithmetic operations, that it uses.

MCS Example

ACME CORP – PROFIT HISTORY

Year	1	2	3	4	5	6	7	8	9
Profit M\$	-3	2	1	-4	5	2	-1	3	-1

Between years 5 and 8 ACME earned

$5 + 2 - 1 + 3 = 9$ Million Dollars

This is the **MAXIMUM** amount that ACME earned in *any* contiguous span of years.

Examples:

Between years 1 and 9 ACME earned

$-3 + 2 + 1 - 4 + 5 + 2 - 1 + 3 - 1 = 4$ M\$

and between years 2 and 6

$2 + 1 - 4 + 5 + 2 = 6$ M\$.

The **Maximum Contiguous Subarray Problem** is to find the span of years in which ACME earned the most, e.g., (5, 8).

Formal Definition

Input: An array of reals $A[1 \dots N]$.

The *value* of subarray $A[i \dots j]$ is

$$V(i, j) = \sum_{x=i}^j A(x).$$

The **Maximum Contiguous subarray problem** is to find $i \leq j$ such that

$$\forall (i', j'), V(i', j') \leq V(i, j).$$

Output: $V(i, j)$ s.t. $\forall (i', j'), V(i', j') \leq V(i, j)$.

Note: Can modify the problem so it returns indices (i, j) .

$\Theta(n^3)$ Solution: Brute Force

Idea: Calculate the value of $V(i, j)$ for each pair $i \leq j$ and return the **maximum** value.

```
VMAX=A[1];
for (i=1 to N) {
  for (j=i to N) {
    // calculate V(i, j)
    V=0;
    for (x= i to j)
      V=V+A[x];
    if (V > VMAX)
      VMAX=V;
  }
}
return VMAX;
```

$\Theta(n^2)$ solution: Reuse data

Idea: We don't need to calculate each $V(i, j)$ from "scratch" but can exploit the fact that

$$V(i, j) = \sum_{x=i}^j A[x] = V(i, j - 1) + A[j].$$

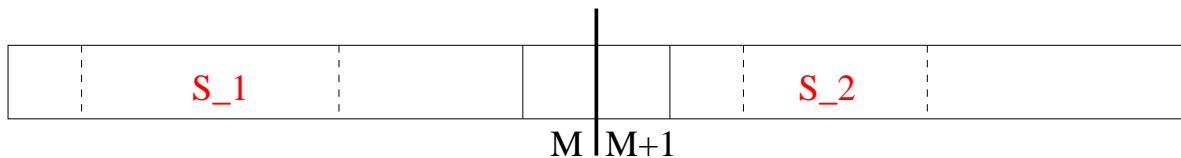
```
VMAX=A[1];
for (i=1 to N) {
  V=0;
  for (j=i to N) {
    // calculate V(i, j)
    V=V+A[j];
    if (V > VMAX)
      VMAX=V;
  }
}
return VMAX;
```

$\Theta(n \log n)$ solution: **Divide-and-Conquer**

Idea: Set $M = \lfloor (N + 1)/2 \rfloor$.

Let A_1 and A_2 be the MCS that MUST contain $A[M]$ and $A[M + 1]$ respectively. Note that the MCS must be one of

- S_1 : The MCS in $A[1 \dots M]$,
- S_2 : The MCS in $A[M + 1 \dots N]$,
- A : Where $A = A_1 \cup A_2$.



$A_1 = \text{MCS on left containing } A[M]$ $A_2 = \text{MCS on right containing } A[M+1]$

$$A = A_1 \cup A_2$$

Example

1	-5	4	2	-7	3	6	-1		2	-4	7	-10	2	6	1	-3
1	-5	4	2	-7	3	6	-1		2	-4	7	-10	2	6	1	-3

$$S_1 = [3, 6] \text{ and } S_2 = [2, 6, 1].$$

$$A_1 = [3, 6, -1] \text{ and } A_2 = [2, -4, 7];$$

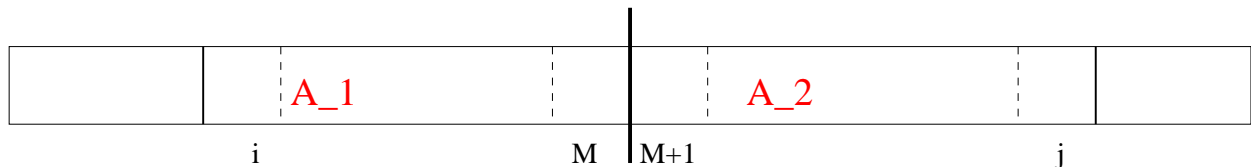
$$A = A_1 \cup A_2 = [3, 6, -1, 2, -4, 7]$$

Since $Value(S_1) = 9$, $Value(S_2) = 9$

and $Value(A) = 13$

the solution to the problem is A .

Finding A : The conquer stage



A_1 is in the form $A[i \dots M]$:

there are only $M - i + 1$ such sequences, so, A_1 can be found in $O(M - i)$ time.

```
MAX=A [M] ;
SUM=A [M] ;
for (k=M-1 down to i)
{
    SUM+=A [k] ;
    if (SUM > MAX) MAX=SUM;
}
A_1=MAX;
```

Similarly, A_2 is in the form $A[M + 1 \dots j]$:

there are only $j - M$ such sequences, so, A_2 the maximum valued such one, can be found in $O(j - M)$ time.

$A = A_1 \cup A_2$ can therefore be found in $O(j - i)$ time, which is linear to the input size.

The Full Divide-and-Conquer Algorithm

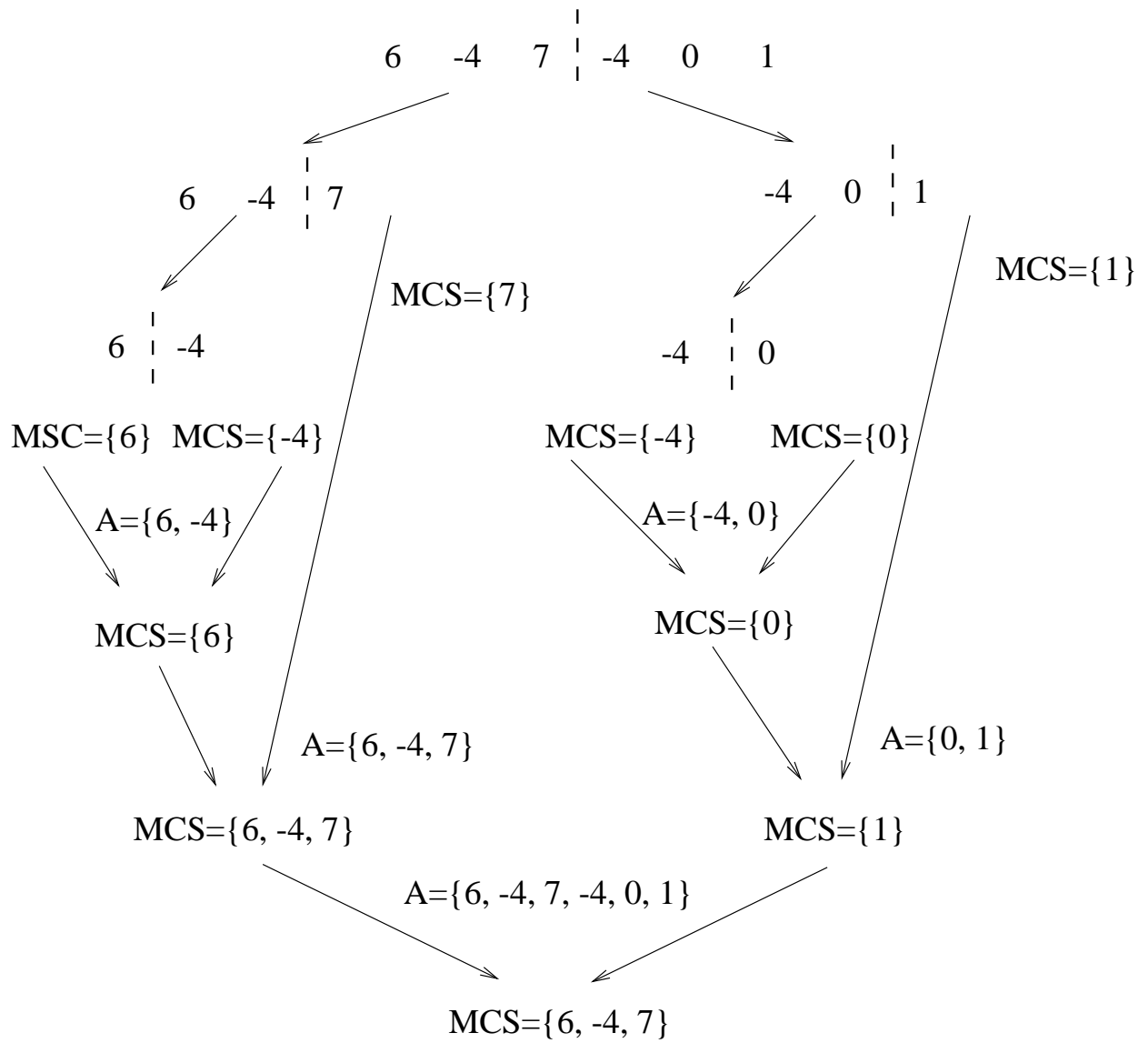
// Input : $A[i \dots j]$ with $i \leq j$

// Output : the MCS of $A[i \dots j]$

$MCS(A, i, j)$

1. If $i == j$ return $A[i]$
2. Else
3. Find $MCS(A, i, \lfloor \frac{i+j}{2} \rfloor)$;
4. Find $MCS(A, \lfloor \frac{i+j}{2} \rfloor + 1, j)$;
5. Find MCS that contains
 both $A[\lfloor \frac{i+j}{2} \rfloor]$ and $A[\lfloor \frac{i+j}{2} \rfloor + 1]$;
6. Return Maximum of the three sequences found

A full example



Analysis of the DC Algorithm

Let $T(m)$ (where m is the problem size) be time needed to run

$$MSC(A, i, j), (j - i + 1 = m)$$

Step (1) requires $O(1)$ time.

Steps (3) and (4) each require $T(m/2)$ time.

Step (5) requires $O(m)$ time.

Step (6) requires $O(1)$ time

Then $T(1) = O(1)$ and

$$\text{for } n > 1, T(n) = 2T(n/2) + O(n)$$

Analysis of the DC Algorithm

To simplify the analysis, we assume that n is a power of 2.

$T(n) \leq 2T\left(\frac{n}{2}\right) + cn$. Repeating this recurrence gives

$$\begin{aligned} T(n) &\leq 2T\left(\frac{n}{2}\right) + cn \\ &\leq 2\left[2T\left(\frac{n}{2^2}\right) + c\frac{n}{2}\right] + cn \\ &= 2^2T\left(\frac{n}{2^2}\right) + 2cn \\ &\leq 2^2\left[2T\left(\frac{n}{2^3}\right) + c\frac{n}{2^2}\right] + 2cn \\ &= 2^3T\left(\frac{n}{2^3}\right) + 3cn \\ &\leq \dots \\ &= 2^hT\left(\frac{n}{2^h}\right) + hcn \end{aligned}$$

Set $h = \log_2 n$, so that $2^h = n$. With this substitution, we have

$$T(n) \leq nT(1) + (\log_2 n)cn = O(n \log_2 n).$$

Review

In this lecture we saw 3 different algorithms for solving the maximum contiguous subarray problem. They were

- A $\Theta(n^3)$ **brute force** algorithm
- A $\Theta(n^2)$ algorithm that **reuses data**.
- A $\Theta(n \log n)$ **divide-and-conquer** algorithm