

CS 381 – FALL 2019

Week 3.3, Friday, Sept 6

Announcement: Homework 1 Solutions released on Piazza
Homework 2 Due: September 16th, 2019 @ 11:59PM on Gradescope

Recap: Master Theorem

- Derived by analyzing recursion tree

$$T(n) = \begin{cases} \mathbf{1000000} & \text{if } n \leq \mathbf{100} \\ a \times T\left(\frac{n}{b} + \mathbf{1}\right) + n^c & \text{otherwise} \end{cases}$$

Obtain Geometric Series: $T(n) = \Theta\left(n^c \sum_{i=1}^{\log_b n} \left(\frac{a}{b^c}\right)^i\right)$

- **Key Ratio:** a/b^c

$$\log_b a \geq c \leftrightarrow \frac{a}{b^c} \geq 1$$

- **Case 1:** $\left(\frac{a}{b^c}\right) < 1$

$$T(n) = \Theta(n^c)$$

- **Case 2:** $\left(\frac{a}{b^c}\right) = 1$

$$T(n) = \Theta(n^c \log n)$$

- **Case 3:** $\left(\frac{a}{b^c}\right) > 1$

$$T(n) = \Theta(n^{\log_b a})$$

Other Form of Master Theorem

- What if MergeCost is not exactly $f(n)=n^c$?
 - $f(n) = n \log n$? or
 - $f(n) = n^{\frac{1}{10} + \log_b a} / \log(n)$?

$$T(n) = \begin{cases} O(1) & \text{if } n \leq 100 \\ a \times T\left(\frac{n}{b} + 50\right) + f(n) & \text{otherwise} \end{cases} \quad \begin{array}{l} \text{Assume} \\ f(n) \geq 0 \end{array}$$

Case 1: $f(n) = \Omega(n^{\varepsilon + \log_b a})$ $T(n) = \Theta(f(n))$

Case 2: $f(n) = \Theta(n^{\log_b a} \log^k n)$ $T(n) = \Theta(n^{\log_b a} \log^{k+1} n)$
(assumes $k \geq 0$)

Case 3: $f(n) = O(n^{\log_b a - \varepsilon})$ $T(n) = \Theta(n^{\log_b a})$

Clicker Question

$$T(n) \leq \begin{cases} 1000000 & \text{if } n \leq 100 \\ a \times T\left(\frac{n}{b} + 50\right) + f(n) & \text{otherwise} \end{cases} \quad \begin{array}{l} \text{Assume} \\ f(n) \geq 0 \end{array}$$

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Case 3: $f(n) = O(n^{\log_b a - \varepsilon})$ $T(n) = \Theta(n^{\log_b a})$

Suppose that $f(n) = n^{\frac{1}{10} + \log_b a} / \log(n)$ above what is $T(n)$?

A. $\Theta(f(n))$

B. $\Theta(n^{\log_b a} \log^{k+1} n)$

C. $T(n) = \Theta(n^{\log_b a})$

D. More info required

Clicker Question

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C. $T(n) = \Theta(n^{\log_b a})$

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Other Types of Recurrences

- $T(n) = T(n - 1) + 1$ (Unroll: $T(n) = \Theta(n)$)
 - $T(n) = T(n - 1) + 1 = T(n - 2) + 1 + 1$
 $= T(n - 3) + 1 + 1 + 1 = \dots = T(n - k) + k$
 $= T(1) + n - 1$

- $T(n) = 2 \times T(n - 10)$ (Exponential)

Two branches

Only constant reduction in input size

$$\begin{aligned} T(n) &= 2T(n - 10) = 2(2T(n - 20)) = 4T(n - 20) \\ &= 8T(n - 30) = \dots = 2^i T(n - 10i) \\ &= 2^{\frac{n}{10}-1} T(10) = \Theta\left(2^{\frac{n}{10}}\right) \end{aligned}$$

Other Recurrences

- $T(n) = T(n - 1) + T(n - 3)$ (Exponential)

Two branches

Only constant reduction in input size

$$T(n) = \Theta(c^n)$$

- Unrolling gets messy fast! How to find c ? [Trick]

Assume $T(n) = c^n$ for some c

$$c^n = T(n) = T(n - 1) + T(n - 3) = c^{n-1} + c^{n-3}$$

$$\rightarrow c^n = c^{n-1} + c^{n-3} \rightarrow c^3 = c^2 + 1$$

$$\rightarrow c \approx 1.46577$$

(Root of Characteristic Equation)

Must verify solution by induction

Other Recurrences

• $T(n) = T(n - 1) + T(n - 3)$ (Exponential)

Two branches Only constant reduction in input size

$$\rightarrow c^n = c^{n-1} + c^{n-3} \rightarrow c^3 = c^2 + 1$$

$$\rightarrow c \approx 1.46577$$

(Root of Characteristic Equation)

Claim: $T(n) \leq kc^n$ (pick k s.t. $T(0) < k$)

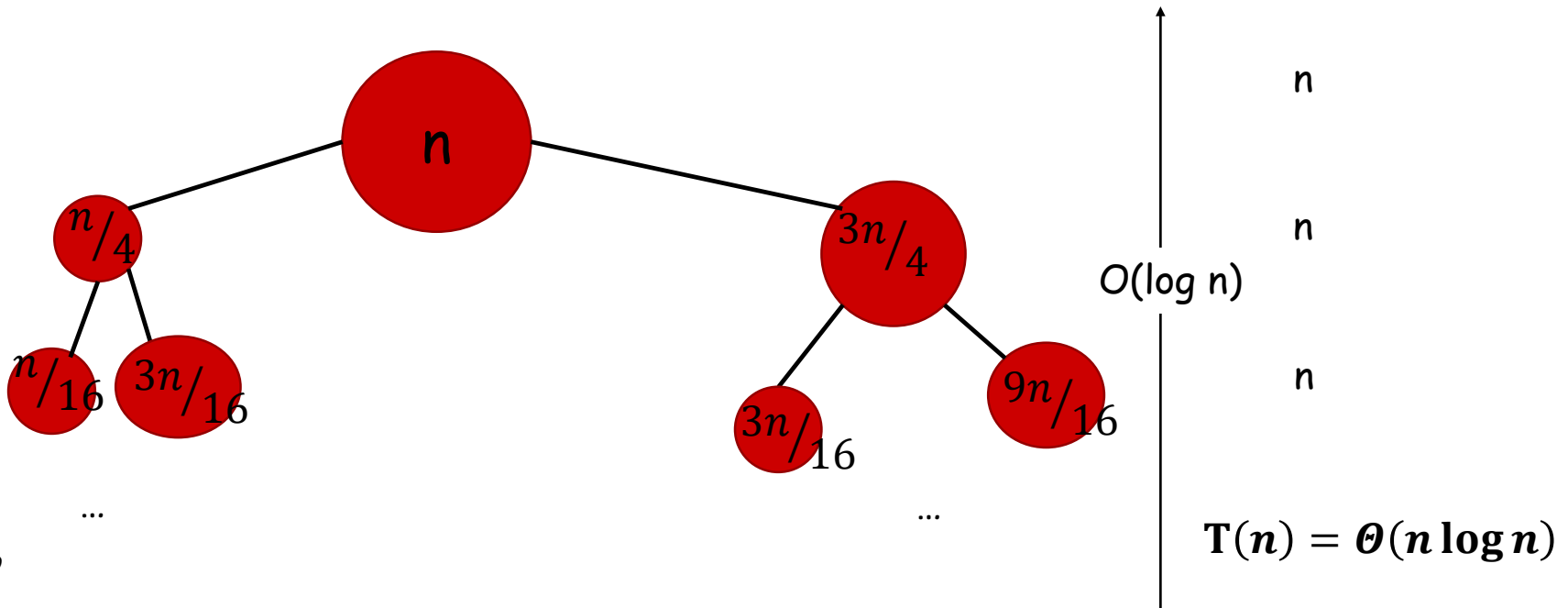
Inductive Step: $T(n) = T(n - 1) + T(n - 3)$
 $\leq k(c^{n-1} + c^{n-3})$ (IH)
 $= k(c^n)$ (Choice of c)

Other Recurrences

- MergeSort with Uneven Split: Split L into A, B of sizes $n/4$ and $3n/4$.

$$T(n) = T\left(\frac{n}{4}\right) + T\left(\frac{3n}{4}\right) + n$$

A bit harder to Analyze with recursion tree

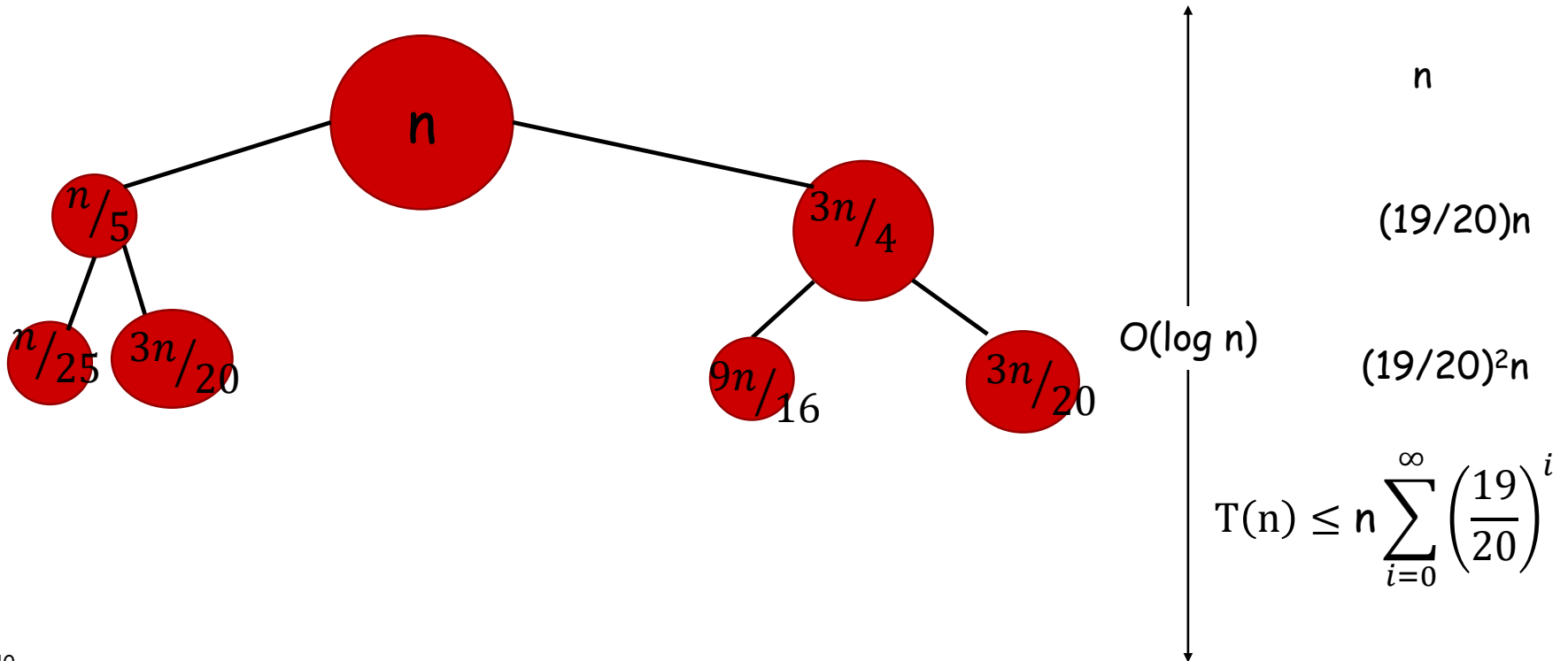


Another Unbalanced Recurrence

- Geometric Series

$$T(n) = T\left(\frac{n}{5}\right) + T\left(\frac{3n}{4}\right) + n$$

- Claim: $T(n) = \Theta(n)$



Divide and conquer algorithms

- ▣ Mergesort
- ▣ Quicksort
- ▣ Binary Search
- ▣ *Linear-time selection*
- ▣ *Skyline Problem*
- ▣ *Maximum Subarray*
- ▣ *Counting inversions*

Maximum Subarray Problem

Given an array A of n (positive and negative) numbers, find the contiguous subarray whose sum has the largest value.

10 5 -20 5 12 -6 33 6 2 -52 6 45 3 -4

Brute Force

- For every pair i and j , $i \leq j$, compute the sum from $A(i)$ to $A(j)$. Remember the pair resulting in the maximum.
- How many pairs? $O(n^2)$
- Using previously computed values, the total running time is $O(n^2)$ (n^3 is excessive brute force)

Aim for $O(n)$ or $O(n \log n)$?

Divide and conquer?

- Split the problem into two halves and solve each recursively
- Combine two solutions to produce the final answer

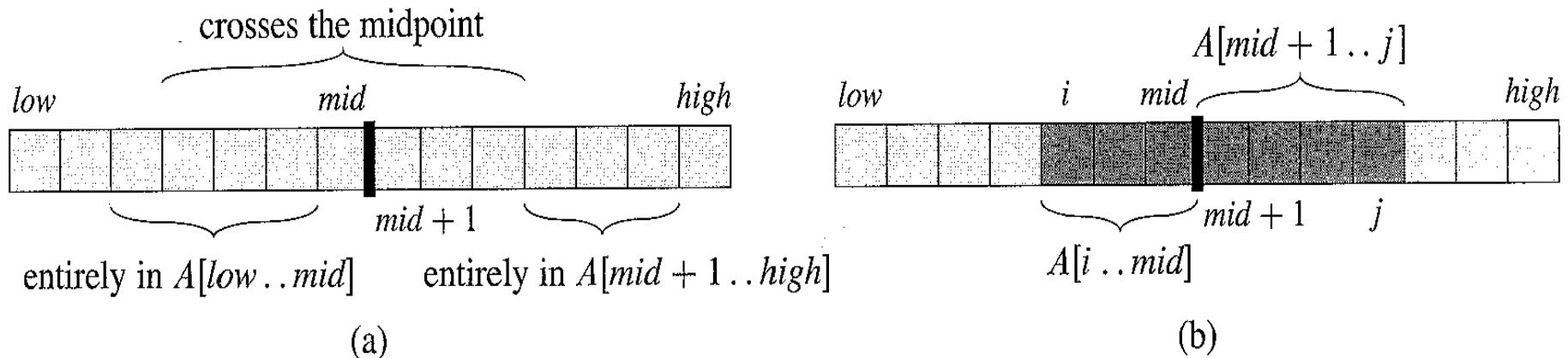


Figure 4.4 (a) Possible locations of subarrays of $A[low..high]$: entirely in $A[low..mid]$, entirely in $A[mid + 1..high]$, or crossing the midpoint mid . (b) Any subarray of $A[low..high]$ crossing the midpoint comprises two subarrays $A[i..mid]$ and $A[mid + 1..j]$, where $low \leq i \leq mid$ and $mid < j \leq high$.

10 5 -20 21 12 -18 12 5 15 -22 6 5 3 -4 -12 55

10 5 -20 21 12 -18 12 5 15 -22 6 5 3 -4 -12 55

10 5 -20 21 12 -18 12 5 15 -22 6 5 3 -4 -12 55



32

46

Combining the answers from the two subproblems

- The maximum subarray can be in one the two halves (easy case) or it can cross the midpoint
- To determine the maximum subarray crossing the midpoint, we can compute the two maximum subarrays “anchored” at the midpoint.
 - Running Sum: sweep left (resp. right) from midpoint
 - keeping track of max
 - Requires time $O(n)$ for merge step

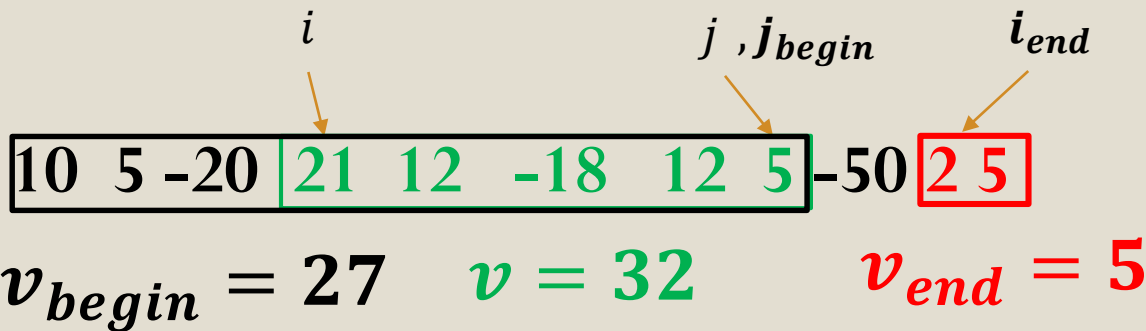
Results in $T(n) = 2T(n/2) + cn$

Gives an **$O(n \log n)$** time algorithm

A Faster $O(n)$ time solution

Idea: Each recursive call returns *additional information* to make merge step easier

- *Optimal Solution:* (i,j) and value $v = \sum_{x=i}^j A[x]$
- *Total Sum:* $T = \sum_{x=1}^n A[x]$
- i_{end} maximizing value $v_{end} = \sum_{x=i_{end}}^n A[x]$
- j_{begin} maximizing value $v_{begin} = \sum_{x=1}^{j_{begin}} A[x]$



A Faster $O(n)$ time solution

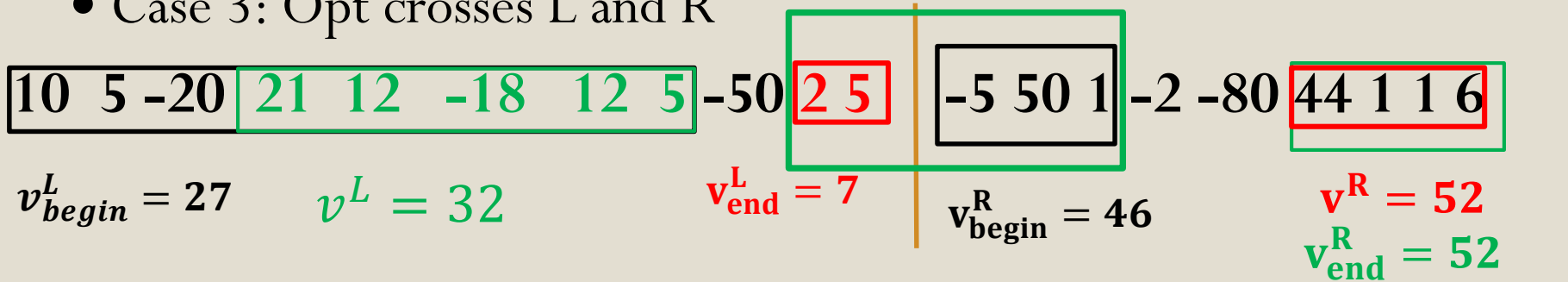
Merge in Constant Time: Suppose A was split into L and R

• Three possibilities for (i, j) : (i^L, j^L) , (i^R, j^R) , (i_{end}^L, j_{begin}^R)

• Case 1: Opt in L

• Case 2: Opt in R

• Case 3: Opt crosses L and R



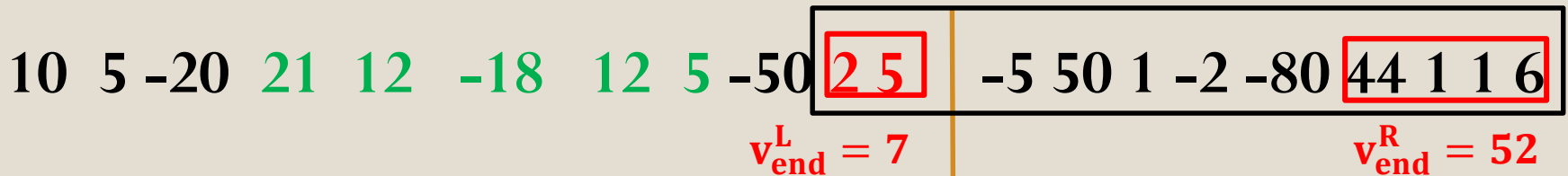
$$v = \max\{v^L, v^R, v_{end}^L + v_{begin}^R\} = v_{end}^L + v_{begin}^R = 53$$

$$\rightarrow (i, j) = (i_{end}^L, j_{begin}^R)$$

A Faster $O(n)$ time solution

Merge in Constant Time: Suppose A was split into L and R

- Still Needs to Compute Extra Values ← For any merges higher in the recursion tree!
- Update Total: $T = T^L + T^R$
- i_{end} maximizing value $v_{end} = \sum_{x=i_{end}}^n A[x]$
 - Case 1: $i_{end} = i_{end}^R$ (interval in R)
 - Case 2: $i_{end} = i_{end}^L$ (interval crosses L)



$$v_{end} = \max\{v_{end}^R, T^R + v_{end}^L\} = v_{end}^R = 52 \rightarrow i_{end} = i_{end}^R$$

A Faster $O(n)$ time solution

Merge in Constant Time: Suppose A was split into L and R

- Merge Needs to Compute Extra Values

- $T = T^L + T^R$

- j_{begin} maximizing value $v_{begin} = \sum_{x=1}^{j_{begin}} A[x]$

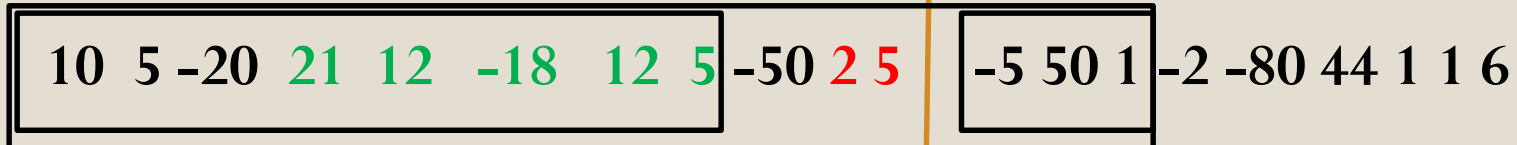
- Similar to computing i_{end}

- Case 1: $j_{begin} = j_{begin}^L$ (interval in L)

- Case 2: $j_{begin} = j_{begin}^R$ (interval crosses R)

$$v_{begin}^L = 27$$

$$v_{begin}^R = 46$$



$$v_{begin} = \max\{v_{begin}^L, T^L + v_{begin}^R\} = T^L + v_{begin}^R = -16 + 46 = 30$$

$$\rightarrow j_{begin} = j_{begin}^R$$

Summary: A Faster $O(n)$ time solution

Idea: Each recursive call returns *additional information* to make merge step easier

Constant Time Merge:

$$T(n) = 2T\left(\frac{n}{2}\right) + 1$$

Master Theorem ($a=2, b=2, c=0$): $T(n) = \Theta(n)$

Summary: Maximum Subarray problem

Given an array A of n (positive and negative) numbers, find the contiguous subarray whose sum has the largest value.

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- *All pairs: $O(n^2)$*
- *D&C with simple combine step: $O(n \log n)$*
- *D&C with extra information: $O(n)$*
- *Simple non-D&C solution: $O(n)$*