

EECS 376 Discussion 3

Sec 27: Th 5:30-6:30 DOW 1017

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Slide deck available at [course drive/Discussion/Slides/Eric Khiu](#)

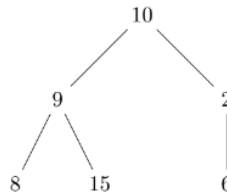
From Last Time:

Binary Tree Local Max Clarification

5. Divide and Conquer

A complete binary tree is a binary tree in which every level, except possibly the last is completely filled and all nodes in the last level are as far left as possible.

Consider a complete binary tree $T = (V, E, r)$ rooted at r where each vertex is labelled with a distinct integer. A vertex $v \in V$ is called a *local maximum* if the label of v is greater than the label of each of its neighbors.



In this example, the local maxima are 10, 15, and 6.

Suppose you are given such a tree where the labelling is given implicitly, i.e., the only way to determine the label of the vertex v is to visit v and query for the vertex label. Provide an algorithm that computes a local maximum of T with using $O(\log(|V|))$ vertex label queries.

Input: Complete, rooted, vertex labelled, binary tree $T = (V, E, r)$

Output: Local maximum v^*

```
1: function COMPUTELOCALMAXIMUM( $T = (V, E, r)$ )
2:   if the label of  $r$  is greater than both of its children's or  $r$  has no children then
3:     return  $r$ 
4:   else
5:     Let  $r'$  be a child of  $r$  with a label greater than  $r$ 
6:     Let  $T'$  be the complete, rooted, vertex labelled, binary tree induced by  $r'$ 
7:     Compute COMPUTELOCALMAXIMUM( $T' = (V', E', r')$ )
```

- ▶ Goal: Return *any* local maximum
- ▶ Q: Is it possible that the tree has multiple local maximums?
 - ▶ Yes, it's possible
- ▶ Q: What if both children are greater than the root?
 - ▶ Recurse into any will work
- ▶ See [Piazza @188](#) for details

Announcement

- ▶ Homework review videos
 - ▶ See [Media Gallery](#) of the Canvas page
 - ▶ Please watch them before submitting regrade request
 - ▶ You should still read the written solutions
 - ▶ Feedback form [here](#)
- ▶ Quick survey: Preference on future discussions:
 - ▶ More content review?
 - ▶ More worksheet problems?
 - ▶ Hard problems review from past homework?

Agenda

- ▶ More on Divide and Conquer
- ▶ Karatsuba Algorithm
- ▶ Dynamic Programming

Divide and Conquer

Lecture Notes

Recap: Divide and Conquer

- ▶ A divide and conquer algorithm usually consists of
 - ▶ Base case(s)
 - ▶ Dividing the problems
 - ▶ Recursive calls
 - ▶ Combining results
- ▶ General form to apply Master Theorem: $T(n) = kT\left(\frac{n}{b}\right) + O(n^d)$
 - ▶ k = number of recursive calls
 - ▶ n/b = size of subproblems
 - ▶ $O(n^d)$ = cost of dividing and combining the results

Correctness Proof

- ▶ Similar idea to prove by induction

Prove by Induction	Correctness Proof for D&C
Prove that $P(0)$ is true	Prove that the base case(s) is/ are correct
Assuming $P(k)$ is true, prove $P(k + 1)$ is true	Assuming recursive calls on smaller inputs return correct answer, prove that the current call is correct Extra: Briefly justify you are making recursive calls under correct <u>condition</u> and with correct <u>input</u>

Worksheet: Array Local Min

1. Array Local Minimum

Let $A[1, \dots, n]$ be an array of distinct integers. The integers in the array are not sorted in any particular order. A cell $A[i]$ is a *local minimum* if $A[i] < A[i - 1]$ and $A[i] < A[i + 1]$. (If $i = 1$ or $i = n$, it only needs to be smaller than the adjacent cell.)

Devise a divide and conquer algorithm to find the local minimum in this array in $O(\log n)$ time.

► Consider

- What are the base case(s)?
- How to divide the problem?
- When to make the recursive calls?
- What is the input to the recursive call?

Worksheet: Array Local Min

1. Array Local Minimum

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Devise a divide and conquer algorithm to find the local minimum in this array in $O(\log n)$ time.

► Consider

- What are the base case(s)? Array size = 1
- How to divide the problem? Start at the middle, recurse into left/ right subarray
- When to make the recursive calls? If at least one of the neighbor is smaller than the middle element of the array
- What is the input to the recursive call?
 - Left subarray if middle > left neighbor
 - Right subarray if middle < right neighbor

Worksheet: Array Local Min- Solution

1. Array Local Minimum

Let $A[1, \dots, n]$ be an array of distinct integers. The integers in the array are not sorted in any particular order. A cell $A[i]$ is a *local minimum* if $A[i] < A[i - 1]$ and $A[i] < A[i + 1]$. (If $i = 1$ or $i = n$, it only needs to be smaller than the adjacent cell.)

Devise a divide and conquer algorithm to find the local minimum in this array in $O(\log n)$ time.

Input: An array $A[1, \dots, n]$ of distinct integers

Output: LOCAL-MIN(A)

```
1: function LOCAL-MIN( $A[1 \dots n]$ )
2:   if  $\text{size}(A) == 1$  then
3:     return  $A[1]$ 
4:    $\text{middle} = \text{size}(A)/2$ 
5:   if  $A[\text{middle}] < A[\text{middle} - 1]$  and  $A[\text{middle}] < A[\text{middle} + 1]$  then
6:     return  $A[\text{middle}]$ 
7:   if  $A[\text{middle}] > A[\text{middle} - 1]$  then
8:     return LOCAL-MIN( $A[1, \dots, \text{middle} - 1]$ )
9:   else
10:    return LOCAL-MIN( $A[\text{middle} + 1, \dots, n]$ )
11:
```

► $T\left(\frac{n}{2}\right) + O(1)$

► Master Theorem says $O(\log n)$

Karatsuba Algorithm

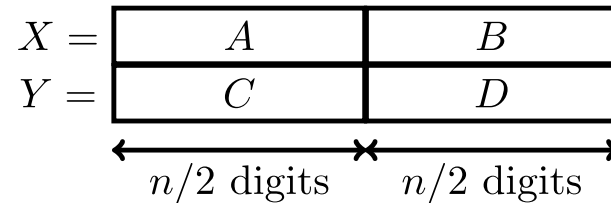
Lecture Notes

Karatsuba Algorithm: Big Idea

- ▶ We want to multiply two numbers X and Y . Each has n digits. Naïve way: $O(n^2)$
- ▶ Rewrite X and Y as follows:

- ▶ $X = A \cdot 10^{\frac{n}{2}} + B$

- ▶ $Y = C \cdot 10^{\frac{n}{2}} + D$



- ▶ Expand $X \cdot Y$ as follows:
 - ▶ $X \cdot Y = \left(A \cdot 10^{\frac{n}{2}} + B \right) \left(C \cdot 10^{\frac{n}{2}} + D \right) = AC \cdot 10^n + (AD + BC) \cdot 10^{\frac{n}{2}} + BD$
- ▶ Observation: The multiplications AC, AD, BC, BD are **smaller versions of the original problem**- we can use Divide and Conquer!
- ▶ Karatsuba Algorithm: Clever “preparations” to make the conquer step faster

Karatsuba Algorithm

- ▶ The Karatsuba Algorithm for Decimal Multiplication is as follows:

```
function KARATSUBA( $x, y$ )  
  //  $n$  here represents the number of digits in the decimal representation of  $x$   
  if  $n = 1$  then  
    return  $x \cdot y$   
  else  
    Write  $x$  as  $a \cdot 10^{n/2} + b$   
    Write  $y$  as  $c \cdot 10^{n/2} + d$   
     $M_1 \leftarrow \text{KARATSUBA}(a, c)$   
     $M_2 \leftarrow \text{KARATSUBA}(b, d)$   
     $M_3 \leftarrow \text{KARATSUBA}(a + b, c + d)$   
    return  $M_1 \cdot 10^n + (M_3 - M_1 - M_2) \cdot 10^{n/2} + M_2$ 
```

Q: What are the
“clever preparations”?

$$M_1 = AC$$

$$M_2 = BD$$

$$M_3 = (A + B)(C + D)$$

- ▶ Remember we wanted $AC \cdot 10^n + (AD + BC) \cdot 10^{n/2} + BD$
 - ▶ Algebra says $M_3 - M_1 - M_2 = AD + BC$
- ▶ Recurrence: $3T\left(\frac{n}{2}\right) + O(n) = O(n^{\log_2 3}) = O(n^{1.585})$
 - ▶ Better than $O(n^2)$

Karatsuba Algorithm: Exercise

- ▶ Compute 37×76
- ▶ $n = 4$ (number of digits)
- ▶ $A = 3, B = 7, C = 7, D = 6$
- ▶ $M_1 = AC = 3 \cdot 7 = 21$
- ▶ $M_2 = BD = 7 \cdot 6 = 42$
- ▶ $M_3 = (A + B)(C + D) = (3 + 7)(7 + 6) = 130$
- ▶ $37 \times 76 = 21 \cdot 10^2 + (130 - 21 - 42) \times 10 + 42 = 2100 + 67 + 42 = 2812$

Algorithm 1 The Karatsuba algorithm for Decimal Multiplication

Input: integers x, y , which are both n -digit numbers with $n \geq 1$

Output: $x \cdot y$

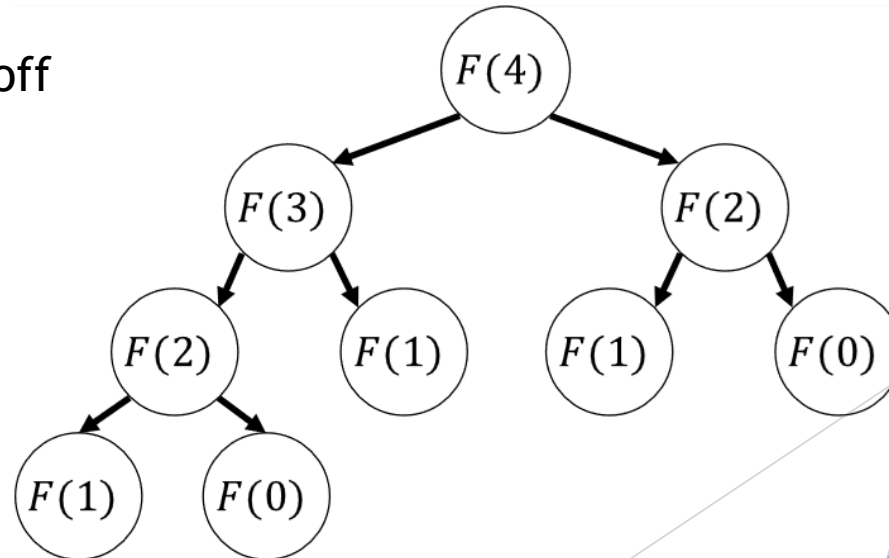
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    if  $n = 1$  then  
        return  $x \cdot y$   
    else  
        Write  $x$  as  $a \cdot 10^{n/2} + b$   
        Write  $y$  as  $c \cdot 10^{n/2} + d$   
         $M_1 \leftarrow \text{KARATSUBA}(a, c)$   
         $M_2 \leftarrow \text{KARATSUBA}(b, d)$   
         $M_3 \leftarrow \text{KARATSUBA}(a + b, c + d)$   
        return  $M_1 \cdot 10^n + (M_3 - M_1 - M_2) \cdot 10^{n/2} + M_2$ 
```

Dynamic Programming

Lecture Notes

Dynamic Programming: Big Idea

- ▶ In D&C, we divide a problem into a smaller versions of the same problem
- ▶ However, for some problems, this recursive subdivision may result in encountering many instances of the **exact same problem**
- ▶ Wouldn't it be nice if we **remember** our solution of duplicated problems so that we **don't have to re-solve them**?
- ▶ Classic debate: Memory-runtime tradeoff
- ▶ In DP, we trade memory for runtime



Divide and Conquer vs DP

Divide and Conquer	Dynamic Programming
Divide original problem to smaller version(s) of the same problem	
Non-overlapping subproblems	Overlapping subproblems
Subproblems usually scale down by a constant: $T(n) \rightarrow T\left(\frac{n}{2}\right) \rightarrow T\left(\frac{n}{4}\right) \rightarrow \dots$	Subproblems don't usually scale down: $T(n) \rightarrow T(n-1) \rightarrow T(n-2) \rightarrow \dots$
Optimal substructure: The solution is correct for this (scaled-down) portion	Optimal substructure: The solution is correct up to this point
Always top-town	Top-down, memoization, bottom-up
Often less time efficient	Often more time efficient- especially with bottom-up

- ▶ Q: Why is MergeSort a D&C algorithm rather than a DP algorithm?
 - ▶ No overlapping subproblems- once a subarray is sorted we never have to sort it again

Useful Notations for Recurrence Relations

- ▶ Big sum/ Big product (Σ , Π)
- ▶ Min/ max (min,max)
- ▶ Argmin/ argmax (argmin,argmax)
- ▶ Big AND/ Big OR (\wedge , \vee)

DP Cookbook

- ▶ Write recurrence
 - ▶ Choose the subject of recurrence
 - ▶ Base case(s)
 - ▶ Form optimal sub-solution (“up to this point”)
- ▶ Size of table (Dimensions? Range of each dimensions?)
- ▶ To fill in cell, which other cells do I look at?
- ▶ Which cell(s) contain the final answer?
- ▶ Reconstructing solution: Follow arrows from final answer to base case

Longest Increasing Subsequence (LIS)

- ▶ Given a sequence of numbers S , an increasing subsequence of S is a subsequence such that the elements are in strictly increasing order.
 - One LIS of $S = [0, 8, 4, 12, 5, 6, 3]$ would be $[0, 4, 5, 6]$
- ▶ Problem: Return the length of LIS of a given sequence
- ▶ Q: Why can't we solve this using divide and conquer?
 - ▶ Overlapping subproblem!

Longest Increasing Subsequence (LIS)

- ▶ Given a sequence of numbers S , an increasing subsequence of S is a subsequence such that the elements are in strictly increasing order.
 - One LIS of $S = [0, 8, 4, 12, 5, 6, 3]$ would be $[0, 4, 5, 6]$
- ▶ Problem: Return the length of LIS of a given sequence
- ▶ **Step 1: Choosing subject of recurrence**
 - ▶ Attempt 1: $L[i]$ = Length of LIS on subarray $S[1, \dots, i]$
 - ▶ Problem: The subsequence chosen can be very ambiguous
 - ▶ Example: In $[0, 8, 4, 12, 5]$, $L[5] = 3$, do we want $[0, 4, 5]$ or $[0, 8, 12]$?
 - ▶ Problematic because when we consider the next element, 6, we can only append it to $[0, 4, 5]$ and have $L[6] = 4$. If my $L[5]$ means $[0, 8, 12]$, then my $L[6]$ is still 3
 - ▶ Reminder: In DP, we don't want to re-solve the subproblem again

Longest Increasing Subsequence

- ▶ Given a sequence of numbers S , an increasing subsequence of S is a subsequence such that the elements are in strictly increasing order.
 - One LIS of $S = [0, 8, 4, 12, 5, 6, 3]$ would be $[0, 4, 5, 6]$
- ▶ Problem: Return the length of LIS of a given sequence
- ▶ **Step 1: Choosing subject of recurrence**
 - ▶ Attempt 2: $L[i]$ = Length LIS on subarray $A[1, \dots, i]$ that **ends at $A[i]$**
 - ▶ Using the same example: $[0, 8, 4, 12, 5]$, $L[5] = 3$
 - ▶ When we consider *any future* element at position k , we **only have to check if it's greater than $L[i]$** to decide if we want to append $S[k]$ to that LIS ending at $S[i]$

Longest Increasing Subsequence

- ▶ Given a sequence of numbers S , an increasing subsequence of S is a subsequence such that the elements are in strictly increasing order.
 - One LIS of $S = [0, 8, 4, 12, 5, 6, 3]$ would be $[0, 4, 5, 6]$
- ▶ Problem: Return the length of LIS of a given sequence
- ▶ **Step 2: Determining base case**
 - ▶ $L[i]$ = Length LIS on subarray $A[1, \dots, i]$ that ends at $A[i]$
 - ▶ Base case: $i = 1$: $L[i] = 1$ (only one element)

Longest Increasing Subsequence

- ▶ Given a sequence of numbers S , an increasing subsequence of S is a subsequence such that the elements are in strictly increasing order.
 - One LIS of $S = [0, 8, 4, 12, 5, 6, 3]$ would be $[0, 4, 5, 6]$
- ▶ Problem: Return the length of LIS of a given sequence
- ▶ **Step 3: Forming optimal sub-solution**
 - ▶ $L[i]$ = Length LIS on subarray $A[1, \dots, i]$ that ends at $A[i]$
 - ▶ (1) We can append $S[i]$ to *any* subsequence ending at $S[j]$, $j < i$
 - ▶ (2) Only append if it is an increasing subsequence, i.e., $S[j] < S[i]$
 - ▶ Which subsequence to append to? **The longest one!** (across all $j < i$)

Longest Increasing Subsequence

- ▶ Given a sequence of numbers S , an increasing subsequence of S is a subsequence such that the elements are in strictly increasing order.
 - One LIS of $S = [0, 8, 4, 12, 5, 6, 3]$ would be $[0, 4, 5, 6]$
- ▶ Problem: Return the length of LIS of a given sequence
- ▶ Putting everything together,

$$L(i) = 1 + \begin{cases} 0 & \text{if } i = 1 \\ \max_{j < i} \{L(j) : S[j] < S[i]\} & \text{otherwise} \end{cases}$$

- ▶ Or equivalently

$$L(i) = \begin{cases} 1 & \text{if } i = 1 \\ 1 + \max_{j < i} \{L(j) : S[j] < S[i]\} & \text{otherwise} \end{cases}$$

Longest Increasing Subsequence

- ▶ Given a sequence of numbers S , an increasing subsequence of S is a subsequence such that the elements are in strictly increasing order.
 - One LIS of $S = [0, 8, 4, 12, 5, 6, 3]$ would be $[0, 4, 5, 6]$
- ▶ Problem: Return the length of LIS of a given sequence
- ▶ Recurrence relation:

$$L(i) = \begin{cases} 1 & \text{if } i = 1 \\ 1 + \max_{j < i} \{L(j) : S[j] < S[i]\} & \text{otherwise} \end{cases}$$

- ▶ What's next? From cookbook:
 - ▶ Size of table (Dimensions? Range of each dimensions?)
 - ▶ To fill in cell, which other cells do I look at?
 - ▶ Which cell(s) contain the final answer?

Longest Increasing Subsequence

- ▶ Given a sequence of numbers S , an increasing subsequence of S is a subsequence such that the elements are in strictly increasing order.
 - One LIS of $S = [0, 8, 4, 12, 5, 6, 3]$ would be $[0, 4, 5, 6]$
- ▶ Problem: Return the length of LIS of a given sequence
- ▶ Recurrence relation:

$$L(i) = \begin{cases} 1 & \text{if } i = 1 \\ 1 + \max_{j < i} \{L(j) : S[j] < S[i]\} & \text{otherwise} \end{cases}$$

- ▶ What's next? From cookbook:
 - ▶ Size of table (Dimensions? Range of each dimensions?) $n \times 1$
 - ▶ To fill in cell, which other cells do I look at? $L[j]$ for all $j < i$
 - ▶ Which cell(s) contain the final answer? $L[n]$

Longest Increasing Subsequence

- ▶ Given a sequence of numbers S , an increasing subsequence of S is a subsequence such that the elements are in strictly increasing order.
 - One LIS of $S = [0, 8, 4, 12, 5, 6, 3]$ would be $[0, 4, 5, 6]$
- ▶ Problem: Return the length of LIS of a given sequence
- ▶ Bottom-up solution:

$$L(i) = \begin{cases} 1 & \text{if } i = 1 \\ 1 + \max_{j < i} \{L(j) : S[j] < S[i]\} & \text{otherwise} \end{cases}$$

Q: How does the max expression translate to algorithm?

Use a variable to keep track of current max

```
LIS( $A[1..n]$ ): // table implementation of LCS  
  allocate  $L[0..n]$   
   $L[0] \leftarrow 0$   
  for  $i = 1..n$ : // fill table  
     $l \leftarrow 0$   
    for  $j = 1..i - 1$ :  
      if  $S[j] < S[i]$ :  $l \leftarrow \max\{l, L[j]\}$   
     $L[i] \leftarrow l + 1$   
  return  $\max_{1 \leq i \leq n} L[i]$ 
```

Runtime: $O(n^2)$

Worksheet: Binary Strings (if time)

Binary Strings

Let $\#C(\ell)$ denote the number of binary strings with length ℓ that have no consecutive occurrences of a 1. For example $\#C(3) = 5$; we can list all binary strings of length 3 and determine by inspection that only the strings 000, 001, 010, 100, and 101 have no consecutive occurrences of a 1.

(c) Show that $\#C(\ell) = \#C(\ell - 1) + \#C(\ell - 2)$ for $\ell \geq 2$. Derive a recursive algorithm to compute $\#C(\ell)$.

Hint: You will need to add base cases.

Solution: Observe that there for a string of length $\ell \geq 2$, it must either begin with a 1 or a 0. If it begins with a 1, then the next bit must be a 0, and the remaining string has length $\ell - 2$ and must have no consecutive occurrences of a 1. If it begins with a 0, then the next bit may be any bit, and the remaining string has length $\ell - 1$ and must have no consecutive occurrences of a 1.

As such, we have that $\#C(\ell) = \#C(\ell - 1) + \#C(\ell - 2)$ for $\ell \geq 2$.

If we want to compute this recursively, we only need to add base cases. Observe that $\#C$ is only defined for non-negative integers, so our base cases should be $\#C(0) = 1$ and $\#C(1) = 2$.

Input: Unsigned integer n

```
1: function  $\#C(n)$ 
2:   if  $n = 0$  then
3:     return 1
4:   if  $n = 1$  then
5:     return 2
6:   return  $\#C(n - 1) + \#C(n - 2)$ 
=0
```

Worksheet: Binary Strings (if time)

Binary Strings

Let $\#C(\ell)$ denote the number of binary strings with length ℓ that have no consecutive occurrences of a 1. For example $\#C(3) = 5$; we can list all binary strings of length 3 and determine by inspection that only the strings 000, 001, 010, 100, and 101 have no consecutive occurrences of a 1.

(d) Use the bottom-up-table approach to improve your recursive algorithm.

Input: Nonnegative integer n

```
1: function  $\#C(n)$ 
2:    $DP \leftarrow$  a table of  $n$  integers
3:    $DP[0] \leftarrow 1$ 
4:    $DP[1] \leftarrow 2$ 
5:   for  $i = 2$  to  $n$  do
6:      $DP[i] \leftarrow DP[i - 1] + DP[i - 2]$ 
7:   return  $DP[n]$ 
```

Next Time

- ▶ More on Dynamic Programming
- ▶ Recursive top-down vs memorization vs bottom up