

CS4800: Algorithms & Data

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Lecture 10:

- Dynamic Programming Wrap-up
- Midterm Review

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Notes for Hackerrank Problems

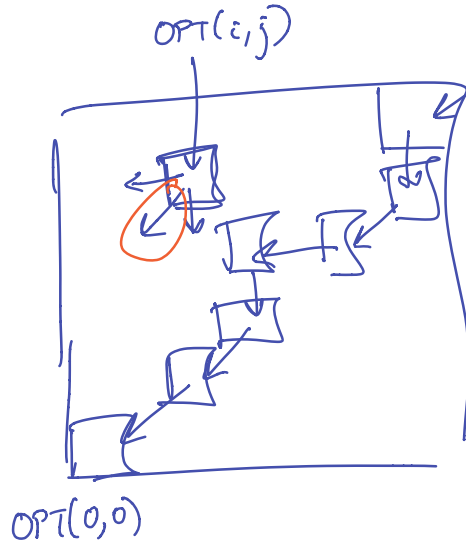
- More public test cases
- List of common issues

Midterm I Review

Edit Distance / Optimal Alignment

x: peas
y: east

p	e	a	s	-
-	e	a	s	t



$OPT(n,m)$

$OPT(i,j)$ is the cost of aligning x_1, \dots, x_i w/ y_1, \dots, y_j

• Each alignment is $n+m$ letters

$$n=m$$

$$n \cdot n \cdot (2n) = 2n^3$$

Analyzing Runtime of DP Algs

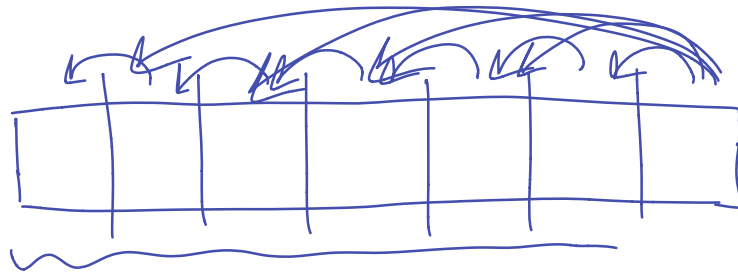
• (Optional: Sorting the input) $O(n \log n)$

- You can use Mergesort w/o proof

- Feel free to "relabel" inputs into sorted order

• Filling the dynamic programming table

- $\leq (\# \text{ of subproblems}) \times (\text{max time to fill each subproblem})$



$$(n-1) \cdot O(1) + O(n) = O(n)$$

• Backtrack through the table to find the solution

- Usually is dominated by the time to fill the table.

Running Time Analysis of Memoization

(Top-Down)

- Let M be a table of solutions to subproblems
- Initialize M to "empty"

Alg(i):

if $M[i]$ not empty : return $M[i]$

else:

$\leq k$ calls [depends on problem]

$M[i] \leftarrow$ [something recursive]

return $M[i]$

Every time we fill one entry of $M[i]$ we make
 $\leq k$ recursive calls. Total # of calls $\leq k \cdot (\# \text{ of entries of } M)$

Running time: $(\# \text{ of calls}) + (\# \text{ of entries of } M) \cdot \left(\begin{array}{l} \text{time to} \\ \text{fill one} \\ \text{entry} \end{array} \right)$

Topics: Induction

- Proof by Induction:

- Mathematical formulae, e.g. $\sum_{i=1}^n i = \frac{n(n+1)}{2}$
- Spot the bug
- Solutions to recurrences
- Correctness of divide-and-conquer algorithms

- Good way to study:

- Lehman-Leighton-Meyer, *Mathematics for CS*
- Review D&C algorithms in Kleinberg-Tardos
- HW2.3

Example Question: Induction

Question: Consider the recurrence $T(n) = T\left(\frac{n}{4}\right) + T\left(\frac{3n}{4}\right) + 2n$

Use the ~~substitution method~~ to show that $T(n) = \Omega(n \log n)$

Solution: We will prove $T(n) \geq cn \log n$ for some c . We first prove the induction step.

$$T(n) = T\left(\frac{n}{4}\right) + T\left(\frac{3n}{4}\right) + 2n \geq \frac{cn}{4} \log \frac{n}{4} + \frac{3cn}{4} \log \left(\frac{3n}{4}\right) + 2n$$

After some math (omitted), we get that $T(n) \geq cn \log n$ is true

whenever $c \leq \frac{2}{2 - \left(\frac{3}{4}\right) \log 3}$. In particular, we can set $c=1$

Topics: Asymptotics

- Asymptotic Notation
 - $o, O, \omega, \Omega, \Theta$
 - Relationships between common function types
- Good way to study:
 - Kleinberg-Tardos Chapter 2

Topics: Asymptotics

Notation	... means ...	Think...	E.g.
$f(n)=O(n)$	$\exists c > 0, n_0 > 0, \forall n \geq n_0:$ $0 \leq f(n) \leq cg(n)$	Upper bound “ \leq ”	$100n^2 = O(n^3)$
$f(n)=\Omega(g(n))$	$\exists c > 0, n_0 > 0, \forall n \geq n_0:$ $0 \leq cg(n) \leq f(n)$	Lower bound “ \geq ”	$2^n = \Omega(n^{100})$
$f(n)=\Theta(g(n))$	$f(n) \in O(g(n))$ and $f(n) \in \Omega(g(n))$	Tight bound “ $=$ ”	$\log(n!) = \Theta(n \log n)$
$f(n)=o(g(n))$	$\forall c > 0, \exists n_0 > 0, \forall n \geq n_0:$ $0 \leq f(n) < cg(n)$	“ $<$ ”	$n^2 = o(2^n)$
$f(n)=\omega(g(n))$	$\forall c > 0, \exists n_0 > 0, \forall n \geq n_0:$ $0 \leq cg(n) < f(n)$	“ $>$ ”	$n^2 = \omega(\log n)$

Topics: Asymptotics

For all real $a > 0$, m , and n , we have the following identities:

$$a^0 = 1,$$

$$a^1 = a,$$

$$a^{-1} = 1/a,$$

$$(a^m)^n = a^{mn},$$

$$(a^m)^n = (a^n)^m,$$

$$a^m a^n = a^{m+n}.$$

For all real $a > 0$, $b > 0$, $c > 0$, and n ,

$$a = b^{\log_b a},$$

$$\log_c(ab) = \log_c a + \log_c b,$$

$$\log_b a^n = n \log_b a,$$

$$\log_b a = \frac{\log_c a}{\log_c b},$$

$$\log_b(1/a) = -\log_b a,$$

$$\log_b a = \frac{1}{\log_a b},$$

$$a^{\log_b c} = c^{\log_b a},$$

where, in each equation above, logarithm bases are not 1.

Topics: Asymptotics

- **Polynomials.** $a_0 + a_1n + \dots + a_d n^d$ is $\Theta(n^d)$ if $a_d > 0$.

$n^3 + 100n^2 + 20n$

- **Logarithms.** $\log_a n = \Theta(\log_b n)$ for all constants $a, b > 0$.

can avoid specifying the base

log grows slower than every polynomial

For every $x > 0$, $\log n = O(n^x)$.

- **Exponentials.** For all $r > 1$ and all $d > 0$, $n^d = O(r^n)$.

Every exponential grows faster than every polynomial

- **Factorial.**

$\log(n!)$

$\log(n!) = \Theta(n \log n)$

factorial grows faster than every exponential

Example Question: Asymptotics

$$2n^2 = o(n^3)$$

Question: Give a value of n_0 which proves the above statement.

Solution: Recall definition

$$\approx \lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = 0$$

$$o(g(n)) = \{f(n) :$$

for any constant $c > 0$,
there is a constant $n_0 > 0$
such that $0 \leq f(n) < cg(n)$
for all $n \geq n_0$ }

Apply definition: $0 \leq 2n^2 < cn^3$

Solve for n : $n > \frac{2}{c}$

$\forall c$ if $n \geq \frac{2}{c} + 1$, then $2n^2 < c \cdot n^3$

Let $n_0 = \frac{2}{c} + 1$

Example Question: Asymptotics

True or False?

1) $n^2 - 5n - 100 = O(n)$

2) $n^3 + 10n^2 + 125 = \omega(n)$

3) $n^2 + O(n) = O(n^2)$

4) $2^{n+1} = O(2^n)$

5) $2^{5n} = O(2^n)$

6) $\log(n^2) = O(\log(n))$

Topics: Recurrences

- Recurrences

- Representing running time by a recurrence
- Solving common recurrences
- Master Theorem

e.g. My alg makes 2 recursive call on inputs of size $n/3$ and does n^2 work

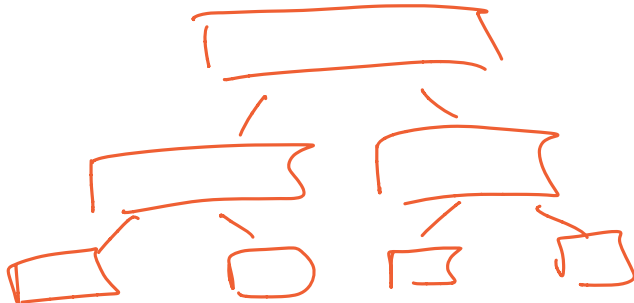
$$T(n) = 2T\left(\frac{n}{3}\right) + n^2$$

- Good way to study:

- Erickson book
- Kleinberg-Tardos D&C Chapter

- Master Theorem

- Recursion Tree



Example Question: Recurrences

Consider the recurrence $T(n) = 5T(n/3) + n$

- a) Draw the recursion tree (at least two levels)
- b) What is the depth of the tree?
- c) What is the amount of work done at level i ?
- d) What is the total amount of work done?

Know your geometric series

$$\sum_{i=0}^n x^i = \frac{x^{n+1} - 1}{x - 1} = \frac{1 - x^{n+1}}{1 - x}$$

Topics: Divide-and-Conquer

- Divide-and-Conquer
 - Writing pseudocode
 - Proving correctness by induction
 - Analyzing running time via recurrences
- Good way to study:
 - Example problems from Kleinberg-Tardos or Erickson
 - HW 2.2, 2.3, 3.1
 - Practice, practice, practice!

Good Review Problems for D+C:

- ① There are n items. In $O(1)$ time you check if item i equals item j . Suppose $> \frac{n}{2}$ of the items are the same. Design an $O(n \log n)$ D+C algorithm to find one such item.
- $O(n \log n)$ algorithm to decide if $> \frac{n}{2}$ items are the same.

- ② n items, each item is either good or bad.

There is a function $HELP(i, j)$ s.t.

$HELP(i, j)$ returns either $\{\text{same, different}\}$

i	j	$HELP(i, j)$	$HELP(i, j) = O(1)$ time
good	good	same	Problem: Suppose $> \frac{n}{2}$ items are good. Find all the good items in $O(n)$.
good	bad	different	
bad	good	different	
bad	bad	same	

Topics: Dynamic Programming

- Dynamic Programming
 - Identify sub-problems
 - Write a recurrence, $OPT(n) = \max\{v_n + OPT(n - 6), OPT(n - 1)\}$
 - Fill the dynamic programming table
 - Find the optimal solution
 - Analyze running time
- Good way to study:
 - Example problems from Kleinberg-Tardos or Erickson
 - HW 4.1, 4.2 (solutions posted Saturday morning)
 - Practice, practice, practice!