

**4.1 – 1 Generalize MATRIX – MULTIPLY – RECURSIVE to multiply  $n \times n$  matrix for which  $n$  is not necessarily an exact power of 2. Give a recurrence describing its running time. Argue that it runs in  $\Theta(n^3)$  time in the worst case**

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MATRIX-MULTIPLY-RECURSIVE( $A, B, C, n$ )
1  if  $n == 1$ 
2  // Base case.
3       $c_{11} = c_{11} + a_{11} \cdot b_{11}$ 
4  return
5  // Divide.
6  partition  $A, B$ , and  $C$  into  $n/2 \times n/2$  submatrices
    $A_{11}, A_{12}, A_{21}, A_{22}; B_{11}, B_{12}, B_{21}, B_{22};$ 
   and  $C_{11}, C_{12}, C_{21}, C_{22};$  respectively
7  // Conquer.
8  MATRIX-MULTIPLY-RECURSIVE( $A_{11}, B_{11}, C_{11}, n/2$ )
9  MATRIX-MULTIPLY-RECURSIVE( $A_{11}, B_{12}, C_{12}, n/2$ )
10 MATRIX-MULTIPLY-RECURSIVE( $A_{21}, B_{11}, C_{21}, n/2$ )
11 MATRIX-MULTIPLY-RECURSIVE( $A_{21}, B_{12}, C_{22}, n/2$ )
12 MATRIX-MULTIPLY-RECURSIVE( $A_{12}, B_{21}, C_{11}, n/2$ )
13 MATRIX-MULTIPLY-RECURSIVE( $A_{12}, B_{22}, C_{12}, n/2$ )
14 MATRIX-MULTIPLY-RECURSIVE( $A_{22}, B_{21}, C_{21}, n/2$ )
15 MATRIX-MULTIPLY-RECURSIVE( $A_{22}, B_{22}, C_{22}, n/2$ )

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This algorithm is based on a matrix of size  $2^b \times 2^b$ , where  $b > 0$  and is an integer

let  $n$  be the size of our matrix, we know  $\exists 2^{b+1}$  greater than  $n$  and  $2^b$  lower than  $n$ . In this case We will construct a  $2^{b+1}$  matrix in which the rows and columns from  $n$  to  $2^{b+1}$  are 0's. We would have:

$$A = \begin{bmatrix} A_{11} & \dots & A_{1n} \\ \vdots & \ddots & \vdots \\ A_{n1} & \dots & A_{nn} \end{bmatrix} \rightarrow \begin{bmatrix} A_{11} & \dots & A_{1n} & \dots & 0 \\ \vdots & \ddots & \vdots & \ddots & \vdots \\ A_{n1} & \dots & A_{nn} & \dots & 0 \\ \vdots & \ddots & \vdots & \ddots & \vdots \\ 0 & \dots & 0 & \dots & 0 \end{bmatrix}$$

$$B = \begin{bmatrix} B_{11} & \dots & B_{1n} \\ \vdots & \ddots & \vdots \\ B_{n1} & \dots & B_{nn} \end{bmatrix} \rightarrow \begin{bmatrix} B_{11} & \dots & B_{1n} & \dots & 0 \\ \vdots & \ddots & \vdots & \ddots & \vdots \\ B_{n1} & \dots & B_{nn} & \dots & 0 \\ \vdots & \ddots & \vdots & \ddots & \vdots \\ 0 & \dots & 0 & \dots & 0 \end{bmatrix}$$

This will be the preprocessing, after it we can call Matrix – Multiple – Recursive

We basically need to create another two arrays of size  $2^{b+1}$  and we know that  $2^b < n < 2^{b+1} < 2n$ . The number elements created are  $2(2^{b+1})^2 < 8n^2$   
 $\rightarrow$  this creation takes  $\Theta(n^2)$

After it we will have a resulting matrix  $C$ , in which the submatrix composed by the first  $n$  rows and the first  $n$  columns are our desired matrix (We will prove it!), all other element  $c_{ij}$  for  $i > n$  or  $j > n$  will be 0

After that We can extract our matrix through  $n^2$  operations.

Our pre processing and postprocessing are  $\Theta(n^2)$ , but our overall algorithm is  $\Theta(n^3)$  so it remains  $\Theta(n^3)$

Now let's prove why this algorithm we created still works

We know that, for  $A$  and  $B$ , without being padded, we have  $c_{ij} = \sum_{k=1}^n a_{ik}b_{kj}$

For our padded  $A$  and  $B$  we have  $c_{ij}^* = \sum_{k=1}^{2^{b+1}} a_{ik}b_{kj}$ , where  $c_{ij}$  are the elements of  $A \cdot B$

and  $c_{ij}^*$  are elements of  $A(\text{padded}) \cdot B(\text{padded})$

we know that  $a_{ik}, b_{ik} = 0 \forall i > n$  and  $k > n$

In this way we have that  $c_{ij}^* = 0 \forall i, \text{ or } j > n$ . Now if we have them  $\leq n$  :

$$c_{ij}^* = \sum_{k=1}^n a_{ik}b_{kj} + \sum_{k=n+1}^{2^{b+1}} a_{ik}b_{kj}, \text{ but for the second summation all we have is } 0$$

$$\text{So we have that } c_{ij}^* = \sum_{k=1}^n a_{ik}b_{kj} = c_{ij} \forall i \text{ and } j \leq n$$

**4.1.2 How quickly can you multiply a  $kn \times n$  matrix by a  $n \times kn$  matrix, where  $k \geq 1$ , using Matrix – Multipl – Recursive as a subroutine?**

**Answer the same question for multiplying  $n \times kn$  by a  $kn \times n$ .**

**Which is Asymptotically better?**

**1<sup>st</sup> case:**

let  $A$  be a  $kn \times n$ . Divide it in  $k$  ( $n \times n$ ) matrices and do the same for  $B$ , the  $n \times kn$  matrix

We now multiply each one of the matrices produced by the division of  $A$  by each one of the matrices produced by the division of  $B$ . This will generate  $k^2 n \times n$  submatrices. After we have each one we can build our resulting matrix

since we will generate  $k^2 n \times n$  submatrices. we will call the subroutine  $k^2$  times

What leads us to a  $\Theta(k^2 n^3)$  running time.

**2<sup>nd</sup> case:**

We do the same. but now we have  $k$  matrices multiplications since we will multiply the  $i^{\text{th}}$  divided submatrix of  $A$  by the  $i^{\text{th}}$  divided submatrix of  $B$ . We will maintain this in memory and sum up all the resulting matrices. What leads us to  $k$  calls of the subroutine

this leads us to  $\Theta(kn^3)$  to end the algorithm properly we need to sum all the  $k$  matrices. each one have  $n^2$  entries, so we add a  $\Theta(kn^2)$ . Resulting in a  $\Theta(kn^3)$ , which is faster

Now we have to prove that the algorithms are correct. We have the following for the first case:

$$A = \begin{bmatrix} A_{11} & \cdots & A_{1n} \\ \vdots & \ddots & \vdots \\ A_{kn,1} & \cdots & A_{kn,n} \end{bmatrix} \text{ and } B = \begin{bmatrix} B_{11} & \cdots & B_{1,kn} \\ \vdots & \ddots & \vdots \\ B_{n1} & \cdots & B_{n,kn} \end{bmatrix}$$

We will divide them into  $k$  submatrices in the form  $n \times n$  as following:

$$\begin{aligned}
A_{k1} &= \begin{bmatrix} A_{11} & \cdots & A_{1n} \\ \vdots & \ddots & \vdots \\ A_{n,1} & \cdots & A_{n,n} \end{bmatrix}; A_{k2} = \begin{bmatrix} A_{n+1,1} & \cdots & A_{n+1,n} \\ \vdots & \ddots & \vdots \\ A_{2n,1} & \cdots & A_{2n,n} \end{bmatrix} \dots \\
A_{kk} &= \begin{bmatrix} A_{(k-1)n+1,1} & \cdots & A_{(k-1)n+1,n} \\ \vdots & \ddots & \vdots \\ A_{kn,1} & \cdots & A_{kn,n} \end{bmatrix} \\
B_{k1} &= \begin{bmatrix} B_{11} & \cdots & B_{1,n} \\ \vdots & \ddots & \vdots \\ B_{n,1} & \cdots & B_{n,n} \end{bmatrix}; B_{k2} = \begin{bmatrix} B_{1,n+1} & \cdots & A_{1,2n} \\ \vdots & \ddots & \vdots \\ B_{n,n+1} & \cdots & A_{n,2n} \end{bmatrix} \dots B_{kk} = \begin{bmatrix} B_{1,(k-1)n+1} & \cdots & B_{1,kn} \\ \vdots & \ddots & \vdots \\ B_{n,(k-1)n+1} & \cdots & B_{n,kn} \end{bmatrix}
\end{aligned}$$

Now we want to show that  $A_{kb} \cdot B_{kj} = \text{submatrix of } A \cdot B \text{ referencing the elements between the rows } (b-1)n+1 \text{ until } bn \text{ and between the columns } (j-1)n+1 \text{ until } jn$

$$= \begin{bmatrix} A_{(b-1)n+1,1} & \cdots & A_{1,n} \\ \vdots & \ddots & \vdots \\ A_{bn,1} & \cdots & A_{n,n} \end{bmatrix} \cdot \begin{bmatrix} B_{1,(j-1)n+1} & \cdots & B_{1,jn} \\ \vdots & \ddots & \vdots \\ B_{n,(j-1)n+1} & \cdots & B_{n,jn} \end{bmatrix}$$

We also know that

$c_{id} = \sum_{h=1}^n a_{ih} b_{hd}$  for the original matrix and  $c_{id}^k = \sum_{h=1}^n a_{ih}^k b_{hd}^k$  for the matrix resulted from the multiplication above. Now, we know that:

$$a_{ih}^k = a_{(b-1)n+i,h} \text{ and } b_{hd}^k = b_{h,(j-1)n+d} \rightarrow$$

$$c_{id}^k = \sum_{h=1}^n a_{(b-1)n+i,h} b_{h,(j-1)n+d} \text{ for } 1 \leq i \leq n \text{ and } 1 \leq d \leq n$$

$$c_{id} = \sum_{h=1}^n a_{ih} b_{hd} \text{ for } 1 \leq i \leq kn \text{ and } 1 \leq d \leq kn.$$

Now lets get the elements in which  $(b-1)n+1 \leq i \leq bn$  and  $(j-1)n+1 \leq d \leq jn \rightarrow$

we can write  $i$  in the form :  $(b-1)n + K, K \text{ from } 1 \text{ to } n.$

we can write  $d$  in the form :  $(j-1)n + D, D \text{ from } 1 \text{ to } n$

We can write the following for all  $k, D \in \mathbb{Z}$ , such  $1 \leq K, D \leq n$

$$c_{(b-1)n+K,(j-1)n+D} = \sum_{h=1}^n a_{(b-1)n+K,h} b_{h,(j-1)n+D} = \sum_{h=1}^n a_{(b-1)n+K,h}^k b_{h,(j-1)n+D}^k = c_{K,D}^k$$

$$c_{(b-1)n+K,(j-1)n+D} = c_{K,D}^k \quad \forall k, D \in \mathbb{Z}, \text{ such } 1 \leq K, D \leq n$$

We can the veracity of the other algoth in a analog manner

**4.1.3 Suppose that instead of partitioning matrices by index calculation in Matrix Algorithm you copy the elements of A, B and C into separate  $n/2 \times \frac{n}{2}$  submatrices.**

**After the recursive calls, you copy the results from  $C_{11}, C_{12}, C_{21}, C_{22}$  back into the appropriate places in C. How does the recurrence change and what is the solution?**

if I make a copy in each iteration, we would have  $3 * n^2$  access to the data, probably multiplied by two because we have to read and write the data in a new place. After that we would need to copy all elements of  $C_{11}, C_{12}, C_{21}$  and  $C_{22}$  back to C. What would take more  $2n^2$

In this case our Recurrence would be  $T(n) = 8 * T\left(\frac{n}{2}\right) + 8n^2$ .

We can use the master theorem and show that this is  $\Theta(n^3)$

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**4.1.4 Write pseudocode for a divide and conquer algorithm MatrixAddRecurse that sums two  $n \times n$  matrices A and B by partitioning each of them into four  $\frac{n}{2} \times \frac{n}{2}$  submatrices and then recursively summing corresponding pairs of submatrices. Assume that matrix partitioning uses  $\Theta(1)$  time index calculations. Write a recurrence for the worst – case running time of this algorithm and solve your recurrence. What happens if you use  $\Theta(n^2)$  time copying to implement the partitioning instead of index calculations?**

We could have:

MATRIX – ADD – RECURSIVE(A, B, C, n)

if  $n == 1$ :

$$c_{11} = c_{11} + b_{11} + a_{11}$$

Copy or index calculation(A, B, C);

MATRIX – ADD – RECURSIVE( $A_{11}, B_{11}, C_{11}, n/2$ )

MATRIX – ADD – RECURSIVE( $A_{12}, B_{12}, C_{12}, n/2$ )

MATRIX – ADD – RECURSIVE( $A_{21}, B_{21}, C_{21}, n/2$ )

MATRIX – ADD – RECURSIVE( $A_{22}, B_{22}, C_{22}, n/2$ )

move  $C_{11}, C_{12}, C_{21}, C_{22}$  to C #if we use the copy method

The recurrence will be  $T(n) = 4T\left(\frac{n}{2}\right) + \Theta(1)$  without copying and

and  $T(n) = 4T\left(\frac{n}{2}\right) + \Theta(n^2)$  with copying

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**4.2.1 Use strassen's algorithm to compute the matrix product:**  $\begin{pmatrix} 1 & 3 \\ 7 & 5 \end{pmatrix} \cdot \begin{pmatrix} 6 & 8 \\ 4 & 2 \end{pmatrix}$

**Strassen's algorithm is based on the idea of reducing the number of multiplications through a very clever way of producing submatrices from the original ones.**

**It's developed to be used on  $2^n \frac{\text{row}}{\text{column}}$  size. for  $n \in \mathbb{Z} > 0$**

*It's just a receipt, we have:*

$$\begin{aligned} S_1 &= B_{12} - B_{22} = 6; S_2 = A_{11} + A_{12} = 4; S_3 = A_{21} + A_{22} = 12; S_4 = B_{21} - B_{11} = -2; \\ S_5 &= A_{11} + A_{22} = 6; S_6 = B_{11} + B_{22} = 8; S_7 = A_{12} - A_{22} = -2; S_8 = B_{21} + B_{22} = 6 \\ S_9 &= A_{11} - A_{21} = -6; S_{10} = B_{11} + B_{12} = 14 \end{aligned}$$

$$\begin{aligned} P_1 &= A_{11} \cdot S_1 = 6 \mid P_2 = S_2 \cdot B_{22} = 8 \mid P_3 = S_3 \cdot B_{11} = 72 \mid P_4 = A_{22} \cdot S_4 = -10 \\ P_5 &= S_5 \cdot S_6 = 48 \mid P_6 = -12 \mid P_7 = S_9 \cdot S_{10} = -84 \end{aligned}$$

$$C_{11} = P_5 + P_4 - P_2 + P_6 = 18$$

$$C_{12} = P_1 + P_2 = 14$$

$$C_{21} = P_3 + P_4 = 62$$

$$C_{22} = P_5 + P_1 - P_3 - P_7 = 66$$

$$C = \begin{pmatrix} 18 & 14 \\ 62 & 66 \end{pmatrix}$$

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**4.2.2 Write pseudocode for Strassen algorithm.**

**STRASSEN( $A, B, C, n$ )**

**if  $n == 1$ :**

$$c_{11} = c_{11} + b_{11} + a_{11}$$

**Initialize all constants  $S_1 \dots S_{10}$**

**Strassen( $A_{11}, S_1, P_1, n/2$ )**

**Strassen( $S_2, B_{22}, P_2, n/2$ )**

**Strassen( $S_3, B_{11}, P_3, n/2$ )**

**Strassen( $A_{22}, S_4, P_4, n/2$ )**

**Strassen( $S_5, S_6, P_5, n/2$ )**

**Strassen( $S_7, S_8, P_6, n/2$ )**

**Strassen( $S_9, S_{10}, P_7, n/2$ )**

$$C_{11} = P_5 + P_4 - P_2 + P_6$$

$$C_{12} = P_1 + P_2$$

$$C_{21} = P_3 + P_4$$

$$C_{22} = P_5 + P_1$$

**Combine ( $C_{11}, C_{12}, C_{21}, C_{22}$ )**

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**4.2.3 What is the largest  $k$  such that if you can multiply  $3 \times 3$  matrices using  $k$  multiplications (not assuming commutativity of multiplication), then you can multiply  $n \times n$  matrices in  $O(n^{\lg 7})$  time? What is the running time of this algorithm?**

Suppose you have a matrix  $n$  that is not a power of 3, you know there exists a constant  $b$  such that:  $3^b < n < 3^{b+1} < 3n$  now you can pad this matrix creating the reminiscent elements.

By this inequality we know this will take  $(3^{b+1})^2 - n^2 \rightarrow n^2 < (3^{b+1})^2 < 9n^2 \rightarrow < 8n^2$

Now we can divide this new size, let say  $k$

$= 3^{b+1}$ , into  $\frac{n}{3} \times \frac{n}{3}$  and we can recurse in the following

$T(n) = k T\left(\frac{n}{3}\right) + O(n^2)$ . if  $n = 3$  then you realize 3 multiplications of size 1.

Using master theorem in  $n^{\log_3 k}$  we have the following situations:

if  $\log_3 k = 2 \rightarrow$  and  $T(n) = \Theta(n^2 \lg n)$ .  $k = 9$  and  $T(n) = O(n^{\lg 7})$

if  $\log_3 k < 2 \rightarrow$  and  $T(n) = \Theta(n^2)$ .  $k < 9$  and  $T(n) = O(n^{\lg 7})$

if  $\log_3 k > 2 \rightarrow$  and  $T(n) = \Theta(n^{\log_3 k})$ .  $k > 9$  and  $T(n) = O(n^{\lg 7})$  when  $k < 3^{\lg 7}$ .  $k$  must be 21

**4.2.4 V. Pan discovered a way of multiplying  $68 \times 68$  matrices using 132,464 multiplications, a way of multiplying  $70 \times 70$  matrices using 143,640 multiplications and a way of multiplying  $72 \times 72$  matrices using 155,424 multiplications. Which method yields the best asymptotic running time when used in a divide-and-conquer matrix multiplication algorithm? How does it compare with Strassen's algorithm?**

**4.2.5 Show how to multiply the complex numbers  $a + bi$  and  $c + di$  using only three multiplications of real numbers. The algorithm should take  $a, b, c$  and as input and produce the real component  $ac - bd$  and the imaginary component  $ad + bc$  separately**

let  $a_1 = ac$

let  $b_1 = bd$

let  $c_1 = (a + b)(c + d)$

the real component is  $a_1 - b_1$ . The imaginary component is  $c_1 - a_1 - b_1$

**4.2.6 Suppose that you have a  $\Theta(n^\alpha)$  time algorithm for squaring a  $n \times n$  matrix where  $\alpha \geq 2$ . Show how to use that algorithm to multiply two different  $n \times n$  matrices in  $\Theta(n^\alpha)$**

Given two matrix  $A \times B$  we want to multiply, create a  $2n \times 2n$  matrix. It have  $4n^2$  elements  
 So it takes  $\Theta(n^2)$  The matrix we want to create is  $\begin{pmatrix} 0 & B \\ A & 0 \end{pmatrix}$ . We then use the algorithm.  
 That takes  $\Theta(2^\alpha n^\alpha)$ . So we have an algorithm that takes  $\Theta(n^\alpha)$

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**4.3.1 Use the substitution method to show that each of the following recurrences defined on the reals has the asymptotic solution specified**

**a).  $T(n) = T(n-1) + n$  has solution  $T(n) = O(n^2)$**

Suppose  $T(k) \leq ck^2 \forall n_0 \leq k < n$

Then  $T(n) = T(n-1) + n \leq c(n-1)^2 + n = cn^2 - 2cn + 1 + n \leq cn^2$

if  $c(1-2n) + n \leq 0$  and if  $n_0 + 1 < n$ . Lets choose  $n_0$  and  $c = 1$  Lets adjust the guess

Suppose  $T(k) \leq k^2 \forall 1 \leq k < n \rightarrow$

$T(n) \leq n^2 - n + 1 \leq n^2$ . If  $1 \leq n-1 \rightarrow 2 \leq n$ . if  $n = 1$   $T(1) = 1 \leq 1$  so

Suppose  $T(k) \leq k^2 \forall 1 \leq k < n$  then we proved that  $T(n) \leq n^2 \forall n \geq 1$  and holds for the base case

**b).  $T(n) = T(n/2) + \Theta(1)$  has solution  $T(n) = O(\lg(n))$**

Suppose  $T(k) \leq clg(k) \forall n_0 \leq k < n$  and  $k = 2^j$  and  $n = 2^l$  for  $j, l \in \mathbb{Z}$  and  $j, l \geq 0$

Then  $T(n) = T\left(\frac{n}{2}\right) + c_1 \leq clg\left(\frac{n}{2}\right) + c_1 = clg(n) - c + c_1$  This if  $\frac{n}{2} \geq n_0 \rightarrow 2n_0 \leq n$  and if  $c = c_1$

Now we need to prove that, given our assumptions that this works for  $n_0 \leq n < 2n_0$

Lets get  $n_0 = 1$ . We need that  $T(1) \leq c_1$ . let  $T(1) = c_1$ . So adjusting our hypothesis

Suppose  $T(k) \leq \Theta(1)lg(k) \forall 1 \leq k < n$  and  $k = 2^j$  and  $n = 2^l$  for  $j, l \in \mathbb{Z}$  and  $j, l \geq 0$

Then  $T(n) \leq \Theta(1)lg(k) \forall 1 \leq n$

**c)  $T(n) = 2T\left(\frac{n}{2}\right) + n$  has solution  $T(n) = \Theta(nlgn)$**

Suppose  $T(k) \leq cklg(k) \forall n_0 \leq k < n$  and  $k = 2^j$  and  $n = 2^l$  for  $j, l \in \mathbb{Z}$  and  $j, l \geq 0$

$T(n) = 2T\left(\frac{n}{2}\right) + n \leq cnlg(n) - cn + n$ . This if  $0 \leq n \geq 2n_0$ . lets take  $n_0 = 2 \rightarrow$

$T(2) = 1 \leq 2c$  if  $c = 1$ . So lets adjust our hypothesis :

Suppose  $T(k) \leq klg(k) \forall 2 \leq k < n$  and  $k = 2^j$  and  $n = 2^l$  for  $j, l \in \mathbb{Z}$  and  $j, l \geq 0$

Then  $T(n) \leq nlgn \forall n \geq 2, n = 2^i$  for  $i \geq 0$   $i \in \mathbb{Z}$ .

We can consider the job to generate a  $2^b$  for any  $n$ . This would take a  $\Theta(n)$ . what wouldn't change  
 The running time

**d)  $T(n) = 2T\left(\frac{n}{2} + 17\right) + n$  has solution  $T(n) = O(n \lg n)$**

Suppose that  $T(k) \leq c k \lg(k) \forall n_0 \leq k < n$  we know that  $\frac{n}{2} + 17 \leq \frac{3n}{4} \forall n \geq 68$

Now  $T(n) = 2T\left(\frac{n}{2} + 17\right) + n \leq 2 \cdot c \cdot \left(\frac{n}{2} + 17\right) \lg\left(\frac{n}{2} + 17\right) + n \rightarrow$

$cn \lg\left(\frac{n}{2} + 17\right) + 34c \lg\left(\frac{n}{2} + 17\right) + n \rightarrow$  if  $n \geq 68 \rightarrow T(n) = 2T\left(\frac{n}{2} + 17\right) + n$

$\leq cn \lg\left(\frac{3n}{4}\right) + 34c \lg(n) + n \rightarrow cn \lg(n) - cn \log\left(\frac{4}{3}\right) + 34c \lg(n) + n \rightarrow$

$T(n) \leq cn \lg(n) + (34c \lg(n) - n \left(\lg\left(\frac{4}{3}\right) - 1\right)) \leq cn \lg(n).$

This all holds if  $n \geq 68$  and  $(34c \lg(n) - n \left(\lg\left(\frac{4}{3}\right) - 1\right)) \leq 0$

$(34c \lg(n) - n \left(\lg\left(\frac{4}{3}\right) - 1\right)) \leq 0 \rightarrow$  if  $c = 20 \rightarrow$

$n > 300$  this holds

Now let's adjust our hypothesis:

Suppose that  $T(k) \leq 20k \lg(k) \forall 300 \leq k < n$  and we know that  $\frac{n}{2} + 17 \leq \frac{3n}{4} \forall n \geq 68$

Now  $T(n) = 2T\left(\frac{n}{2} + 17\right) + n \leq 2 \cdot 20 \cdot \left(\frac{n}{2} + 17\right) \lg\left(\frac{n}{2} + 17\right) + n \rightarrow$  if  $\left(\frac{n}{2} + 17\right) \geq 300$

$T(n) \leq 20n \lg(n)$  if  $(n \geq 566)$  What if  $300 \leq n$

$< 566$ ? We know that the algorithm always terminate

So we have a set  $F\{T(n); \text{for } 300 \leq n < 566\}$

by the well ordering principle we know it has a maximum, let say  $c_{\text{special}} = \max(F)$ .

Clearly  $T(n) \leq c_{\text{special}} \leq c_{\text{special}} n \lg(n)$ . if  $c_{\text{special}} \leq 20$  we are done. if  $c_{\text{special}}$

$\geq 20$  we will use  $c$  as  $c_{\text{special}}$ . In this case we the  $n$  we choosed early can still be considered so we are done

**e)  $T(n) = 2T\left(\frac{n}{3}\right) + \Theta(n)$  has solution  $T(n) = \Theta(n)$**

Suppose  $T(k) \leq 3c_2 k \forall n_0 \leq k < n$  where  $k$  is  $3^i$  for some  $i \in \mathbb{Z}$  and  $i > 0$  Then:

We know that  $\exists c_1, n_0^*$  and  $c_2$  we used above such that  $0 \leq c_1 n \leq \Theta(n) \leq c_2 n \forall n \geq n_0 \rightarrow$

$T(n) = 2T\left(\frac{n}{3}\right) + \Theta(n) \leq \frac{2 \cdot 3c_2 n}{3} + \Theta(n) \leq 2c_2 n + c_2 n \rightarrow$

$T(n) \leq 2c_2 n$  if  $n_0 \leq \left(\frac{n}{3}\right)$ . We have to see the cases  $n_0 \leq n$

$< 3n_0$ . One more time they are finite and have a maximum

if  $c \geq$  maximum, we are done. if not, get  $c$  as the maximum. This proves  $O(n)$

The proof for  $\Omega$  is analagous, using min and the lower bound property



f)  $T(n) = 4T\left(\frac{n}{2}\right) + \Theta(n)$  has solution  $T(n) = \Theta(n^2)$

Suppose  $T(k) \leq ck^2 - \frac{c_1}{3}k$  for  $n_0 \leq k < n \rightarrow$

$$T(n) = 4T\left(\frac{n}{2}\right) + \Theta(n) \leq 4c\left(\frac{n}{2}\right)^2 + c_1n - \frac{4}{3}c_1n \leq cn^2 - \frac{1}{3}c_1n \text{ but this is valid only if } \frac{n}{2} > n_0$$

We need to verify for  $n_0 \leq n < 2n_0$ .

Again we can choose  $c = \max\{T(n); \text{for } n_0 \leq n < 2n_0\}$

The proof for lower bound is similar

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**4.3.2 The solution to the recurrence  $T(n) = 4T\left(\frac{n}{2}\right) + n$  turns out to be**

**$T(n) = \Theta(n^2)$ . Show that a substitution proof with  $T(n) \leq cn^2$  fails.**

**Then show how to subtract a lower order to make a substitution proof work**

Suppose  $T(k) \leq ck^2$  for  $n_0 \leq k < n \rightarrow$

$$T(n) = 4T\left(\frac{n}{2}\right) + n \leq cn^2 + n \text{ which isn't guaranteed that } T(n) \leq cn^2$$

$$\text{Now let's subtract a lower term and we have } T(n) \leq cn^2 - nc_1 - nc_1 + n \\ = cn^2 - nc_1 - n(c_1 - 1) \leq cn^2 - nc_1 \text{ if } c_1 \geq 1$$

We can do a similar analysis for the lower bound:

---

**4.3.3 The recurrence  $T(n) = 2T(n-1) + 1$  has the solution  $T(n) = O(2^n)$**

**Show that a substitution proof fails with the assumption  $T(n) \leq c2^n$  is constant**

**Then show how to subtract a lower order term to make a substitution proof work**

Suppose  $T(k) \leq c2^k$  for  $n_0 \leq k < n$ . Then we would have:

$$T(n) = 2T(n-1) + 1 \leq 2 \cdot c \cdot 2^{n-1} + 1 = c \cdot 2^n + 1$$

We can subtract a lower term,  $d$ , what leads us to:

$$T(n) = 2T(n-1) + 1 \leq 2 \cdot c \cdot 2^{n-1} (-2d + 1)$$

If we choose  $-2d + 1 \leq 0$  we are done

---

**4.4.1 For each of the following recurrences, sketch its recursion tree, and guess a good asymptotic upper bound on its solution. Then use the substitution method to verify your answer**

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

$$(n)^3 \rightarrow \left(\frac{n}{2}\right)^3 \rightarrow \left(\frac{n}{4}\right)^3 \rightarrow \dots \rightarrow 1$$

There are  $\lg(n) + 1$  of size. The sum of the work done on each node is

$$\sum_{k=0}^{\lg(n)} \left(\frac{n}{2^k}\right)^3 = \sum_{k=0}^{\lg(n)} \frac{n^3}{8^k} = n^3 \sum_{k=0}^{\lg(n)} \frac{1}{8^k} = S.$$

$$\text{Now } S - \frac{1}{8}S = \sum_{k=0}^{\lg(n)} \frac{1}{8^k} - \frac{1}{8^{k+1}} = \frac{7S}{8} = \left(1 - \frac{1}{8^{\lg(n)+1}}\right)n^3 \rightarrow S = \frac{1}{7}\left(8 - \frac{1}{8^{\lg(n)}}\right)n^3 \leq \frac{8}{7}n^3$$

This could give us a better upper bound, but we will guess its only cubic by now

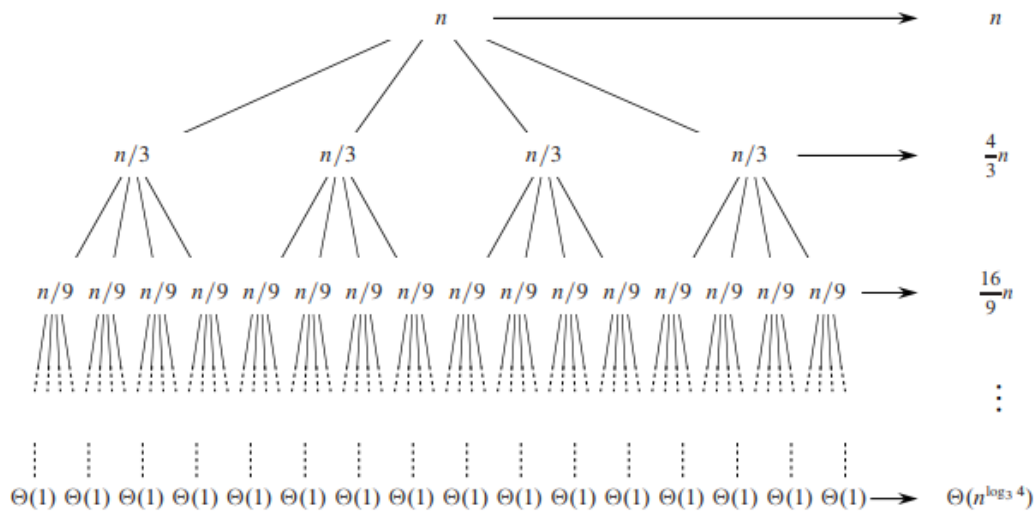
Suppose  $T(k) \leq ck^3$  for  $n_0 \leq k < n$  then:

$$T(n) = T\left(\frac{n}{2}\right) + n^3 \leq \left(\frac{c}{8}\right)^1 n^3 + n^3 = \left(1 + \frac{c}{8}\right)n^3. \text{ Now we just need } c \text{ such } 1 + \frac{c}{8} \leq c$$

This is valid for  $\frac{n}{2} \geq n_0$ , we need to verify for  $n_0 \leq n$

$< 2n_0$ . We then use the same argument as before, get the max of  $T(n)$  and use as the  $c$

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



The height is  $\log_3(n)$

$$S = \sum_{k=0}^{\log_3 n} n \left(\frac{4}{3}\right)^k \rightarrow \frac{4}{3}S = \sum_{k=0}^{\log_3 n} n \left(\frac{4}{3}\right)^{k+1} \rightarrow \frac{1}{3}S = n \sum_{k=0}^{\log_3 n} \left(\frac{4}{3}\right)^{k+1} - \left(\frac{4}{3}\right)^k = \left(\frac{4}{3}^{\log_3 n+1} - 1\right)n$$

$$S \leq 3n \left(\frac{4}{3}\right)^{\log_3 n+1} \leq 4 \cdot 4^{\log_3 n} \rightarrow S \leq 4n^{\log_3 4}$$

Now let's prove it: Suppose  $T(k) \leq c_1 k^{\log_3 4} - c_2 n$  for  $n_0 \leq k < n$

$$T(n) = 4 \cdot T\left(\frac{n}{3}\right) + n \leq \frac{4c_1 n^{\log_3 4}}{4} + n - \frac{4}{3}c_2 n \leq c_1 n^{\log_3 4} \text{ if } 1 - \frac{4}{3}c_2 < -c_2$$

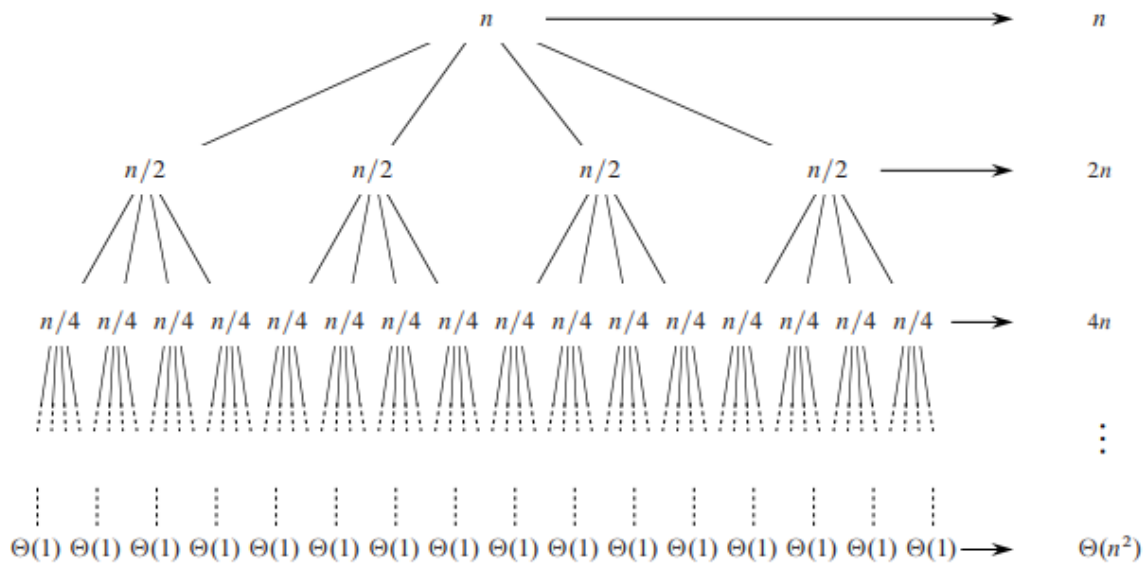
get  $c_2 > 3$ . But this is valid only for  $n > 3n_0$ .

Now get the  $c = \max(T(n))$  for  $n_0 \leq n < 3n_0$ .

$$T(n) \leq cn^{\log_3 4}. \text{ Now we can take a } c_3 \text{ such } T(n) \leq c_3 n^{\log_3 4} \leq cn^{\log_3 4} - c_2 n$$

Basing on the last resolutions QED

c)  $T(n) = 4T\left(\frac{n}{2}\right) + n \rightarrow$



The size of the tree is  $\log_2 n + 1$  the Running time can be estimated as

$$S = \sum_{k=0}^{\lg(n)} 2^k n \rightarrow 2S - S = S = \sum_{k=0}^{\lg(n)} 2^{k+1} n - 2^k n = (2^{\lg_2 n + 1} - 1)n = 2n^2 - n$$

Our guess is that  $T(n) = O(n^2)$  So Suppose  $T(k) \leq c_1 k^2 - c_2 k$  for  $n_0 \leq k < n$  then:

And choose  $c_2$  such  $4c_2 - 1 = c_2 \rightarrow 3c_2 = 1 \rightarrow c_2 = \frac{1}{3}$

$$T(n) = 4 \cdot T\left(\frac{n}{2}\right) + n \leq 4 \cdot c_1 \left(\frac{n}{2}\right)^2 + n - 4c_2 n = c_1 n^2 - n(4c_2 - 1) = c_1 n^2 - c_2 n$$

if  $n_0 \leq n < 2n_0$ . Taking  $n_0 = 2$  we have  $2 \leq n < 4 \rightarrow$  let's analyse for  $n = 2$  and  $n = 3$

$$T(2) = 1 \leq 4c_1 - \frac{2}{3} \text{ and } T(3) = 1 \leq 9c_1 - 1. \text{ Just take } c_1 = 1 \text{ and we are done}$$

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



The height clearly will be  $n$  and our guess  $S = \sum_{k=1}^n 3^k \rightarrow 2S = 3^{n+1} - 1 \rightarrow$  our guess is

$$O(3^n). \text{ Suppose } T(k) \leq c_1 3^k - c_2 \text{ for } n_0 \leq k < n \rightarrow T(n) = 3T(n-1) + 1$$

$$\leq 3c_1 3^{n-1} - 3c_2 + 1 = c_1 3^n - c_2 + (1 - 2c_2) \leq c_1 3^n - c_2 \left( \text{choose } c_2 = \frac{1}{2} \right)$$

But this is valid for  $n_0 \leq n-1 \rightarrow n \geq n_0 + 1$ . But for  $T(n_0) = 1$

$$< c_1 3^{n_0} - c_2. \text{ Choose } n_0 = 2 \text{ and a } c_1 \text{ and } c_1 = 1 \rightarrow T(2) = 1 \leq 9 - 1/2$$

**4.4.2 Use the substitution method to prove that recurrence 4.15 has the asymptotic lower bound  $L(n) = \Omega(n)$ . Conclude that  $L(n) = \Theta(n)$**

$$4.15: L(n) = \begin{cases} 1, & \text{if } (n < n_0) \\ L\left(\frac{n}{3}\right) + L\left(\frac{2n}{3}\right) & \end{cases}$$

Suppose  $c_1 k \leq L(k)$  for  $n_0^* \leq k < n$ , take  $n_0 = 5n_0^*$

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

$c_1 n = c_1 \frac{n}{3} + c_1 \frac{2n}{3} \leq L(n)$  for  $5n_0 \leq \frac{n}{3} \rightarrow$  for  $n \geq 15n_0$  let analyze for

$5n_0 \leq n < 15n_0$  we know that this will be bounded and we take the  $x = \{\min \text{ of } T(n) / \max \text{ of } n; 5n_0 \leq n < 15n_0\}$

now take the  $z = \min x, c_1$  and we know that:

$zn \leq L(n)$  for  $5n_0 \leq n < 15n_0$  and we are done

**4.4.3 Use the substitution method to prove the recurrence 4.14 has the solution  $T(n) = \Omega(n \lg(n))$ . Conclude  $T(n) = \Theta(n \lg n)$**

$$4.14 : T(n) = T\left(\frac{n}{3}\right) + T\left(\frac{2n}{3}\right) + \Theta(n)$$

Lets prove the lower bound first. We know exist  $n_0$  such that  $\Theta(n) \leq c_1 n \forall n \geq n_0$

Now lets suppose:  $T(k) \leq c k \lg(k) - c_3 k$  for  $n_0 \leq k < n$  then:

$$T(n) = T\left(\frac{n}{3}\right) + T\left(\frac{2n}{3}\right) + \Theta(n) \leq c \frac{n}{3} \lg\left(\frac{n}{3}\right) + c \frac{2n}{3} \lg\left(\frac{2n}{3}\right) + c_2 n - \frac{c_3 n}{3} - \frac{2c_3 n}{3}:$$

$$cn \left( \frac{1}{3} \lg\left(\frac{n}{3}\right) + \frac{2}{3} \lg\left(\frac{2n}{3}\right) \right) - n(c_2 - c_3) = cn \lg \left( \left(\frac{n}{3}\right)^{\frac{1}{3}} \left(\frac{2n}{3}\right)^{\frac{2}{3}} \right) - n(c_2 - c_3)$$

$$cn \lg \left( \frac{n^{\frac{1}{3}} 2^{\frac{2}{3}} n^{\frac{2}{3}}}{3^{\frac{1}{3}} 3^{\frac{2}{3}}} \right) - n(c_2 - c_3) = cn \lg \left( \frac{n 2^{\frac{2}{3}}}{3^1} \right) - n(c_2 - c_3) \leq cn \lg(n) - n(c_2 - c_3)$$

Now lets choose  $c_3 = \frac{c_2}{2} \rightarrow T(n) \leq cn \lg(n) - nc_2$  if  $n \geq 3n_0$  if  $n_0 \leq n < 3n_0$ .

let  $c_{\max} = 100 * \max\{T(n); n_0 \leq n < 3n_0\}, c, c_2\} \rightarrow T(n) \leq c_{\max} \leq c_{\max} n \lg(n) - c_2 n$

so  $T(n) = O(n \lg(n))$ . We make an analogous proof for the lower bound :

Suppose  $ck \lg(k) \leq T(k)$  for  $0 \leq k < n$

$$c \frac{n}{3} \lg\left(\frac{n}{3}\right) + c \frac{2n}{3} \lg\left(\frac{2n}{3}\right) \leq c \frac{n}{3} \lg\left(\frac{n}{3}\right) + c \frac{2n}{3} \lg\left(\frac{2n}{3}\right) + c_2 n \leq T(n)$$

$$c \frac{n}{3} \lg\left(\frac{n}{3}\right) + c \frac{2n}{3} \lg\left(\frac{2n}{3}\right) + c_2 n = cn \left( \lg\left(\frac{n 2^{\frac{2}{3}}}{3^1}\right) = \lg(n) + \lg\left(2^{\frac{2}{3}}\right) - \lg(3) \right) \leq T(n)$$

$$cn \lg(n) \leq cn \lg(n) + \frac{2}{3} cn - cn \lg(3) + c_2 n \leq T(n) \text{ if } \frac{2}{3} cn - cn \lg(3) + c_2 n \geq 0$$

$$c \left( \frac{2}{3} - \lg(3) \right) \geq -c_2 \rightarrow c \left( \lg(3) - \frac{2}{3} \right) \leq c_2 \rightarrow c \leq \frac{c_2}{\lg(3) - \frac{2}{3}} \text{ and we are done for}$$

$n_0 \leq n < 3n_0$  We now get the min of  $T(n) / \max n$  and  $c$  so  $T(n) < c * n \lg(n)$

**4.4.4 Use a recursion tree to justify a good guess for the solution to the recurrence  $T(n) = T(\alpha n) + T((1 - \alpha)n) + \Theta(n)$ , where  $\alpha$  is a constant such  $0 < \alpha < 1$**

We know that the work done on the childrens are equivalent to the work done on the parent node now we need to find the size of it. Without loss of generality, take  $\alpha > 1 - \alpha$ .

In this case we will have the deepest leaf at the node in which  $n\alpha^{\text{height}-1} = 1 \rightarrow$

$(\text{height} - 1) = \log_a n^{-1} \rightarrow \text{height} = \log_a \left(\frac{1}{n}\right) + 1$ . We do  $\Theta(n)$  in each level So

$$\begin{aligned} (\log_a a - \log_a n)\Theta(n) &\rightarrow \left(\log_a \frac{a}{n}\right)cn \rightarrow \frac{\left(\log_{\frac{1}{a}} \frac{a}{n}\right)}{\log_{\frac{1}{a}} a} = \frac{\left(\log_{\frac{1}{a}} a - \log_{\frac{1}{a}} n\right)}{\log_{\frac{1}{a}} a} = \frac{-1 - \log_{\frac{1}{a}} n}{-1} \\ &= \left(1 + \log_{\frac{1}{a}} n\right)cn \text{ we will assume its bounded by } n \lg(n) \end{aligned}$$

Suppose  $T(k) \leq c k \lg(k)$  for  $n_0 \leq k < n$

$$T(n) = T(\alpha n) + T((1 - \alpha)n) + \Theta(n) \leq$$

$$T(n) \leq c(\alpha n) \log \alpha n + c(1 - \alpha)n \lg((1 - \alpha)n) + c_2 n$$

$$\leq cn[\log(\alpha^\alpha n^\alpha) + \log(1 - \alpha)n^{(1-\alpha)}] + cn$$

$$T(n) \leq cn[\alpha \log(\alpha) + \alpha \log(n) + (1 - \alpha) \log(n) + (1 - \alpha) \log(1 - \alpha)] + c_2 n$$

$$T(n) \leq cn[\log(n) + \alpha \log(\alpha) + \log(1 - \alpha) \log(1 - \alpha)] + c_2 n \rightarrow$$

$$T(n) \leq cn \log(n) + n[c[\log(1 - \alpha) \log(1 - \alpha) + \alpha \log(\alpha) + c_2]] \text{ this is true if we have:}$$

$$[c[\log(1 - \alpha) \log(1 - \alpha) + \alpha \log(\alpha) + c_2]] \leq 0 \rightarrow$$

$$c \geq -\frac{c_2}{[\log(1 - \alpha) \log(1 - \alpha) + \alpha \log(\alpha)]} \text{ This value here is a constant, so get it!}$$

Now to hold we know that  $0 \leq \Theta(n) \leq c_2 n$  for  $n$

$\geq n_0$ . We will get this  $n_0$  be the same as our hypothesis  $n_0$

We showed by far that if  $T(k) \leq c k \lg(k)$  for  $n_0 \leq k < n$  then  $T(n) \leq c n \lg n$  if

$\alpha n \geq n_0$  and  $(1 - \alpha)n \geq n_0$  get  $n^*$  as  $\min\{\alpha, (1 - \alpha)\}$  then we proved for

$n_0 \leq n < n^* n_0$  But for the same analysis we did previously we can get the minimum

of the running times, compare with the constant  $c$ , find minimums and maximums

and choose the values that guarantess that it all works

**4.5.1 Use the master method to give tight asymptotic bounds for the following recurrences**

a)  $T(n) = 2T\left(\frac{n}{4}\right) + 1$

$$a = 2; b = 4, f(n) = 1. 1 = O(n^{\log_4 2^{-E}}). \text{ take } E = \frac{1}{3} \text{ and this holds } T(n) \text{ is } \Theta\left(n^{\frac{1}{2}}\right)$$

b)  $T(n) = 2T\left(\frac{n}{4}\right) + \sqrt{n}$

$$a = 2; b = 4, f(n) = \sqrt{n} \text{ There is no } E \text{ such that } n^{\frac{1}{2}} = O\left(n^{\frac{1}{2}-E}\right) \text{ since every } \epsilon \text{ would}$$

make the exponent of  $g(n) = n^{\frac{1}{2}-E}$ , less than  $\frac{1}{2}$ . We can prove it by contradiction.

Suppose that exists such an  $E$ . Then  $\exists c_1$  and  $n_0$  such that  $\forall n \geq n_0$

$$0 \leq n^{\frac{1}{2}} \leq c_1 n^{\frac{1}{2}-E} \rightarrow \text{this would imply that } n^E \leq c \text{ for all } n \geq n_0 \text{ now take } n > \frac{1}{c^E}$$

This would imply  $n^E > c$ . Thus a contradiction.

Lets analyze the second case of the Master Theorem:

There exists a constant  $K \geq 0$  such that  $n^{\frac{1}{2}} = \Theta\left(n^{\frac{1}{2}} \lg^k n\right)$ ? Yes  $k = 0$

This leads to  $T(n) = \Theta(n^{\frac{1}{2}} \lg(n))$

$$c) T(n) = 2T\left(\frac{n}{4}\right) + \sqrt{n} \lg^2 n$$

$a = 2; b = 4, f(n) = \sqrt{n} \lg^2 n$  There exists  $k \geq 0$  such  $n^{\frac{1}{2}} \lg^2 n = \Theta\left(n^{\frac{1}{2}} \lg^k n\right), k = 2$

Then  $T(n) = \Theta(n^{\frac{1}{2}} \lg^3 n)$

$$d) T(n) = 2T\left(\frac{n}{4}\right) + n$$

$f(n) = n \rightarrow n = \Omega\left(n^{\frac{1}{2} + \frac{1}{2}}\right)$  (Here we take  $E = \frac{1}{2}$ ) Additionally  $2f\left(\frac{n}{4}\right) \leq cf(n) \rightarrow$   
 $\frac{n}{2} \leq cn$  for  $c = \frac{2}{3}$  Then  $T(n) = \Theta(n)$

$$e) T(n) = 2T\left(\frac{n}{4}\right) + n^2$$

$f(n) = n^2 \rightarrow n^2 = \Omega\left(n^{\frac{1}{2} + \frac{3}{2}}\right)$  (Here we take  $E = \frac{3}{2}$ ) Additionally  $2f\left(\frac{n}{4}\right) \leq cf(n) \rightarrow$   
 $\frac{n^2}{8} \leq cn^2$  for  $c = \frac{2}{3}$  Then  $T(n) = \Theta(n^2)$

**4.5.2 Professor Caesar wants to develop a matrix multiplication algorithm that is asymptotically faster than Strassen algorithm. His algorithm will use the divide and conquer methodm dividing each matrix intro  $\frac{n}{4} \times \frac{n}{4}$  submatrices and the divide and combine steps together will take  $\Theta(n^2)$  time. Suppose that the professor Algorithm creates a recursive subproblems of size  $\frac{n}{4}$ . What is the lasgest integer value of  $a$**

We need largest  $a$  such that  $\log_4 a < \lg 7 \rightarrow a = 48$

**4.5.3 Use the master method to show that th solution to the binary search recurrence  $T(n) = T\left(\frac{n}{2}\right) + \Theta(1)$  is such that  $T(n) = \Theta(\lg(n))$**

$a = 1, b = 2, f(n) = 1$ . Exists  $e$  such that  $1$

$= O(n^{-E})$ ? No we could prove it, but its obviously false

for case 2 exists  $k \geq$  such that  $1 = \Theta(1)$ . take  $k = 0 \rightarrow \Theta(\lg(n))$ ;

**4.5.4 Consider the function  $f(n) = \lg(n)$ . Argue that although  $f\left(\frac{n}{2}\right) < f(n)$ , the regularity condition  $af\left(\frac{n}{b}\right) \leq cf(n)$  with  $a = 1$  and  $b = 2$  does not hold for any constant  $c < 1$ . Argue further that for any  $E > 0$ , the condition in case 3 that  $f(n) = \Omega(n^{\log_b a + E})$  does not hold**

$$\lg\left(\frac{n}{2}\right) \leq c \lg(n) \rightarrow \lg(n) - 1 \leq c \lg(n) \rightarrow (1 - c) \lg(n) \leq 1.$$

This clearly does not hold since  $\lg(n)$  is unbounded

Suppose it holds for some  $E$ , then  $0 \leq c_1 n^E \leq \lg(n)$ . Then  $\frac{n^E}{\lg(n)} \leq c_2 \forall n \geq n_0 \rightarrow$

But  $\lim_{n \rightarrow \infty} \frac{n^E}{\lg(n)} = \lim_{n \rightarrow \infty} \frac{\ln(2)En^E}{1} = \infty$ . Thus is not bounded above. So there is no  $E$ .

**4.5.5 Show that for suitable constants  $a, b$  and  $E$ , the function  $f(n) = 2^{\lceil \lg(n) \rceil}$  satisfies all the conditions in case 3 of the master theorem except the regularity condition**

Lets begin by showing that the regularity doesn't hold for any  $a, b, c \rightarrow$

$$a2^{\lceil \lg(\frac{n}{b}) \rceil} \leq a2^{\lceil \lg(\frac{n}{b}) \rceil} \leq 2a2^{\lg(\frac{n}{b})}$$

$$a2^{\lg(\frac{n}{b})} = \frac{a}{b} > 2c = 2c2^{\lg(n)} \geq$$

By Inspection we find that the numbers  $a = 1$   $b = \sqrt{2}$  and  $c = 1$  guarantees the case 3. Now lets use this numbers in the regularity problem

$$a2^{\lceil \lg(\frac{n}{b}) \rceil} = c2^{\lceil \lg n \rceil} \rightarrow 2^{\lceil \lg n - \frac{1}{2} \rceil} = c2^{\lceil \lg n \rceil}, \text{ but exists } n \text{ such that } n = 2^k \text{ for } k \in \mathbb{Z} > 0 \rightarrow$$

$$2^{\lceil \lg 2^k - \frac{1}{2} \rceil} = 2^{\lceil \lg 2^k - \frac{1}{2} \rceil} \text{ but } \left\lceil k - \frac{1}{2} \right\rceil = [k] \text{ so } c \text{ would need to be}$$

$> 1$  for this case. So doesn't hold

#### 4.1 Recurrence examples

**Give asymptotic tight upper and lower bounds for  $T(n)$  in each of the following algorithmic recurrences. Justify your answers**

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

We will guess our  $\Theta(g(n))$  based on the tree method. in the first line we have  $n^3$

In the second line we have  $\frac{n^3}{4}$ , in the third line we have  $4 \frac{n^3}{4^3} = \frac{n^3}{16}$ .

So in each level we can guess it will be  $\frac{n^3}{4^{i-1}}$ . We expect that  $\frac{n}{2^{\text{height}-1}}$  is 1 in the last leaves

$n = 2^{\text{height}-1} \rightarrow \log_2 n + 1 = \text{height}$ . So we expect to have the following  $T(n)$

$$T(n) = \sum_{k=0}^{\log_2 n} \frac{n^3}{4^k} \rightarrow S = n^3 \sum_{k=0}^{\log_2 n} \frac{1}{4^k} \rightarrow \frac{S}{4} = n^3 \sum_{k=0}^{\log_2 n} \frac{1}{4^{k+1}} \rightarrow S - \frac{S}{4} = \sum_{k=0}^{\log_2 n} \frac{1}{4^k} - \frac{1}{4^{k+1}} \rightarrow$$

$$\frac{3S}{4} = 1 - \frac{1}{4^{\log_2(n)+1}} \rightarrow S = \frac{4}{3} \left( 1 - \frac{1}{(2^2)^{\log_2(n)+1}} \right) = S = \frac{4}{3} \left( 1 - \frac{1}{(2^{\log_2(n)+1})^2} \right)$$

$$S = \frac{4}{3} n^3 \left( 1 - \frac{1}{(2^{\log_2(n)} \cdot 2)^2} \right) = \frac{4n^3}{3} \left( 1 - \frac{1}{(2n)^2} \right) \rightarrow S = \frac{4n^3}{3} - \frac{n}{3}$$

So we will guess that  $T(n) = \Theta(n^3)$

Suppose that  $T(k) \leq c_1 k^3$  for  $n_0 \leq k < n$

Then  $T(n) = 2T\left(\frac{n}{2}\right) + n^3 = \frac{2c_1 n^3}{8} + n^3 = n^3 \left( \frac{1c_1}{4} + 1 \right)$  Now we choose  $c_1$  such

$\frac{c_1}{4} + 1 = c_1 \rightarrow \frac{3c_1}{4} = 1 \rightarrow c_1 = \frac{4}{3}$ . but this is valid for  $2n_0 \leq n$ . We need to verify for

$n_0 \leq n < 2n_0$ . Lets choose  $n_0 = 2$ . Then  $T(1) = 1 \leq \frac{4}{3}$ ;  $T(2) = 1 \leq \frac{32}{3}$ ;

$T(3) = 2 + 27[T(1) + T(2) + n^3] \leq \frac{108}{3}$ .

So Assuming that  $T(k) \leq \frac{4}{3} k^3$  for  $2 \leq k < n$  we can prove that  $T(n)$

$\leq \frac{4}{3} n^3$ . By the strong inductive argument this is true for 2

$\leq n$ . Since its true for  $T(2)$

This proves that  $T(n) = O(n^3)$ . Now lets prove that it is  $\Omega(n^3)$ :

Suppose that  $\frac{4}{3} k^3 \leq T(k)$  for  $n_0 \leq k < n$  then we would have:

$T(n) = 2T\left(\frac{n}{2}\right) + n^3 \geq \frac{2 \cdot \frac{4}{3} n^3}{8} + n^3 = \frac{n^3}{3} + n^3 = \frac{4}{3} n^3$ . If  $n \geq 2n_0$  We need to verify for  $n_0 \leq n < 2n_0$ . Now lets choose  $n_0 = 2$  again. In this case ne need to verify for 2 and 3

$T(2) = 1 \leq \frac{4 \cdot 8}{3}$ . Take  $c_2$  equals  $\frac{1}{100}$ . Then  $T(2) = 1 \geq \frac{8}{100}$  and  $T(3) = 1 \geq \frac{27}{100}$

In addition since  $c_2$  we choosed first was greater than our final  $c_2$  our derivation remains valid In this case lets rearragen our hypothesis, what leads us to

Suppose that  $\frac{1}{100} k^3 \leq T(k)$  for  $2 \leq k$

$< n$  then this would imply that this is valid for  $n$

Since its true for  $T(2)$  by the strong inductive argument we can guarantee that this holds for all  $n \geq 2$ . So this is  $\Omega(n^3)$ . Since it is  $O(n^3)$  we can conclude that is  $\Theta(n^3)$

**b)  $T(n) = T\left(\frac{8n}{11}\right) + n$**

Lets use the master theorem in this case. We have  $a = 1$ ,  $b = \frac{11}{8}$ , and  $f(n) = n$

Lets analyze case 1; Does exist a  $E > 0$  such that  $n = O(n^{-E})$ . No. Suppose exists.

Then would have the following situation:

$\exists c_1$  and  $n_0 > 0$  such that  $\forall n \geq n_0$  we would have  $n \leq c_1 n^{-e}$

$\rightarrow$  this would imply that the  $g(n) = \frac{n}{n^{-e}} = n^{1+e}$  is bounded.

Which is clearly unbounded. So lets see case 2:

Does exist  $K \geq 0$  such that  $n$

$= \Theta(\lg^k n)$ ? Obvisouly no. Lets prove that there is no such  $k$

Assume there is such  $K$ , then we would have positive constants  $c_1 c_2$  and  $n_0$  that

$\forall n \geq n_0$  we would have  $0 \leq c_1 \lg^k(n) \leq n \leq c_2 \lg^k(n)$ . This would imply that the



Function  $\frac{n}{lg^k(n)}$  is bounded. But if we use L'hospital in the limit:

$$\lim_{n \rightarrow \infty} \frac{n}{lg^k(n)} = \frac{\ln(2)}{k} \cdot \frac{x}{\log^{k-1}(n)} = \left( \frac{\ln^2 2}{k \cdot (k-1)} \right) \cdot \frac{x}{\log^{k-2}(n)} \dots \text{After } k \text{ derivatives,}$$

$$\text{we could prove the formula by induction on } k \rightarrow = \lim_{n \rightarrow \infty} \frac{\ln^k 2}{k!} x$$

$= \infty$  So its clearly unbounded. So there is no such  $K > 0$

Lets analyse the last case Does exist a constant  $E$  such that  $n$

$$= \Omega(n^E) \text{ and if } f(n) \text{ additionally satisfies af } \left(\frac{n}{b}\right)$$

$$\leq cf(n) \text{ for some constant } c < 1 \text{ and } n \geq n_0?$$

Take  $E = 1$  and clearly  $n = \Omega(n)$  Now  $f\left(\frac{8}{11}n\right) = \frac{8}{11}n \leq cn$ . Take  $c$

$$= \frac{9}{11} \text{ and holds for all } n.$$

Last case Holds, so we have  $T(n) = \Theta(n)$

$$\text{c) } T(n) = 16T\left(\frac{n}{4}\right) + n^2$$

Lets use the master theorem again  $a = 16, b = 4, f(n) = n^2$ .

Case 1. There exists  $e > 0$  such that  $f(n) = O(n^{\log_b a - e})$ ;  $n^2$

$$= O(n^{2-e})? \text{ Clearly no. if so, we would have that } n^e \text{ is bounded above.}$$

Case 2. There exists  $K \geq 0$  such that  $f(n) = \Theta(n^{\log_b a} \log^k n)$ ? Yes  $k = 0$

So we have that  $T(n) = \Theta(n^2 \log(n))$

$$\text{D) } T(n) = 4T\left(\frac{n}{2}\right) = n^2 \log(n)$$

We will use the master theorem  $a = 4, b = 2, \log_b a = 2$ , and  $f(n) = n^2 \lg(n)$

case 1, there exists  $e > 0$  such that  $f(n) = O(n^{2-E})$ ?  $n^2 \lg(n)$

$$= O(n^{2-E}) \text{ Clearly not. If we had such } E > 0, \text{ this would lead us that}$$

$n^E \lg(n)$  is bounded, which is clearly false.

Case 2, there exists  $k \geq 0$  such that  $n^2 \lg(n) = \Theta(n^2 \lg^k n)$ . Yes, take  $k = 1$ , Then

$$T(n) = \Theta(n^2 \log^2 n)$$

$$\text{E) } T(n) = 8T\left(\frac{n}{3}\right) + n^2$$

$$a = 8, b = 3, \log_b a = \log_3 8, f(n) = n^2$$

Case 1 There exists a constant such that  $n^2 = O(n^{\log_3 8 - E})$ ? No.  $\log_3 8 - E$

$$< 2. \text{ this would imply that } n^l \text{ for } l$$

$$> 0 \text{ is bounded above, which is clearly false}$$

Case 3. There exists a constant  $E > 0$  such that  $n^2 = \Omega(n^{\log_3 8 + \epsilon})$ ? Yes  $\log_3 8 < 2$ . So

Exist  $E > 0$  such that  $\log_3 8 + k = 2$  which leads us to  $n^2 \Omega(n^2)$  lets analyze regularity

$$8f\left(\frac{n}{3}\right) = \frac{8}{9}n^2 \leq cn^2? \text{ take } c = \frac{8.1}{9}. \text{ So we have that } T(n) = \Theta(n^2)$$

Case 2 There exists a constant such that  $n^2 = \Theta(n^{\log_3 8} \lg^k n)$ ?

$$F) T(n) = 7T\left(\frac{n}{2}\right) + n^2 \lg(n)$$

$$a = 7, b = 2, f(n) = n^2 \lg(n).$$

Case 1, there exists  $E > 0$  such that  $n^2 \lg(n) = O(n^{\log_2 7 - E})$ ? yes

Lets find such  $c_1$  and  $n_0$  such that  $\forall n \geq n_0$  we have

$$0 \leq n^2 \lg(n) \leq c_1 n^{\log_2 7 - E}. \text{ we need such an } E \text{ that makes}$$

$$\lim_{n \rightarrow \infty} n^{2+E-\log_2 7} \lg(n) \leq c_1. \text{ We know that } \log_2 7 \leq 2.800001 \rightarrow$$

$$\lim_{n \rightarrow \infty} n^{-0.8} \lg(n) \rightarrow \frac{\lg(n)}{n^{0.8}} = 0. \text{ There is such an } E. \text{ Then we have } T(n) = \Theta(n^{\log_2 7})$$

$$G) T(n) = 2T\left(\frac{n}{4}\right) + \sqrt{n}$$

Case 1: There exists  $E > 0$  such that  $n^{\frac{1}{2}}$

$$= O\left(n^{\frac{1}{2}-E}\right)? \text{ No, this would imply that } n^E \text{ is bounded above}$$

Case 3: There exists  $E > 0$  such that  $n^{\frac{1}{2}}$

$$= \Omega\left(n^{\frac{1}{2}+E}\right). \text{ No. This would imply that } n^E \text{ is bounded above.}$$

Case 2: There exists  $K \geq$  Such that  $n^{\frac{1}{2}} = \Theta\left(n^{\frac{1}{2} \lg^k n}\right)$ ? Yes take  $k = 0$ . In this case we have

$$\text{In this case we have } T(n) = \Theta\left(n^{\frac{1}{2} \lg(n)}\right)$$

$$H) T(n) = T(n-2) + n^2$$

Lets guess a bound through the bound method. We have  $n^2$  done in the first,  $(n-2)^2$  in the second,  $(n-4)^2$  in the third. Lets guess the height.

$$\text{We have that } (n - 2(\text{height} - 1)) = 1 \rightarrow \frac{n+1}{2} = \text{height}$$

$$\text{We will guess that } T(n) = \Theta(n^3)$$

Suppose that  $T(k) \leq c_1 k^3$  for  $n_0 \leq k < n$  then we would have:

$$T(n) = T(n-2) + n^2 \leq c_1(n-2)^3 + n^2 \rightarrow$$

$$T(n) \leq c_1 n^3 - 6c_1 n^2 + 12nc_1 + n^2 \rightarrow \text{we want } -6c_1 n^2 + 12nc_1 + n^2$$

$$< 0 \text{ we want } 1 - 6c_1 < 0 \rightarrow c_1 > \frac{1}{6} \text{ guarantess that it goes to } -\infty$$

$T(n) \leq c_1 n^3$  if  $n-2 \geq n_0$  if  $n \geq n_0 + 2 \rightarrow$  if  $n_0 \leq n < n+2$ . We need to analyze

Since the limit above goes to infinity exists such  $n_0$  that guarantees it. We want now:

$$T(n_0) = 1 \leq c_1 n_0^3 \rightarrow \text{choose } c_1 = 1 \text{ and } n_0 = \max\{\text{previous } n_0 \text{ and } 1\}$$

$$T(n_0 + 1) = 1 \leq T(n_0) \leq c_1(n_0 + 1)^3. \text{ This proves that its Upper bound is } O(n^3).$$

We can do a similar analysis for the lower bound, we would have similarly:

$$c_2 k^3 \leq T(k) \text{ if } n_0 \leq k < n \text{ then}$$

$$c_2 n^3 \leq c_2 n^3 - 6c_2 n^2 + 12nc_2 + n^2 \leq T(K) \rightarrow \text{This happens if } -6c_2 n^2 + 12nc_2 + n^2$$

$$> 0 \text{ for } n \geq n_0. \text{ Just get } c_2 \text{ such that } 1 - 6c_2 \geq 0 \text{ } c_2 \leq \frac{1}{6}$$

Just analyze the  $n_0$  necessary for the quadratic equation above be positive always and analyze

The values for  $n_0 \leq n < n_0 +$

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#### 4.2 Parameter – passing costs

Throughout this book, we assume that parameter passing during procedure calls takes constant time, even if – element array is being passed.

This assumption is valid in most systems because a pointer to the array is passed, not the array itself, This problem examines The implications of three parameter – passing strategies:

1. Arrays are passed by pointer. Time =  $\Theta(1)$

2. Arrays are passed by copying. Time =  $\Theta(N)$ , where  $N$  is the size of the array

3. Arrays are passed by copying only the subrange that might be accessed by the called procedure. Time =  $\Theta(n)$  if the subarray contains  $n$  elements

Consider the following htree algorithms:

The recursive binary search algorithm for finding a number in a sorted array.

The Merge – Sort procedure

Give six recurrences  $T_{a1}(N, n) \dots T_{b3}$  for the worst case running time of each of the three algorithms above when arrays and matrices are passed using each of the three parameter – passing strategies above. Solve your recurrences giving tight asymptotic bounds

**Recursive Binary:**

**Arrays are passed by pointer**

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

Now lets use the master theorem  $a = 1, b = 2, f(n) = \Theta(1)$

does exist  $e > 0$  such that  $\Theta(1) = O(n^{-e})$ . Clearly no, since the limit of  $n^{-e}$  is 0

Does exist  $k \geq 0$  such that  $\Theta(1) = \Theta(\lg^k n)$ ? Yes.  $\Theta(1) = \Theta(1)$ . So  $T(n) = \Theta(\lg(n))$

**Arrays are passed by copying**

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

Lets make a guess through the tree method. In the first node we have  $\Theta(N)$

In the second we have  $\Theta(N) \dots$  The height of our tree is such that  $\frac{N}{2^{h-1}} = 1 \rightarrow$

$$\log(N) + 1 = h \rightarrow \text{Then we have } \sum_{h=1}^{\log N + 1} \Theta(N) = (\log(N) + 1)\Theta(N)$$

We will then assume it is  $\log(n)$ . Now lets prove it by strong induction

Suppose that  $T(k) \leq c_1 k \lg(k)$  for  $n_0 \leq k < N$

$$\text{Then } T(n) = T\left(\frac{n}{2}\right) + \Theta(N) \leq c_1 \frac{N}{2} \lg\left(\frac{N}{2}\right) + c_2 N = c_1 \frac{N}{2} \log(N) - c_1 \frac{N}{2} + c_2 N \leq$$

$$c_1 N \log(N) - N \left(\frac{c_1}{2} - c_2\right) \leq c_1 N \log(N). \text{ This if } \frac{c_1}{2} - c_2 \leq 0 \rightarrow$$

Just take  $c_1 \leq 2c_2$ . and  $n_0$  such that  $0 \leq \Theta(N) \leq c_2 N \forall n \geq n_0$

Ok but this to happen we need that  $\frac{n}{2} \geq n_0$ . we need to verify if it holds for

$n_0 \leq n < 2n_0$  By the same argument as in previous exercises  $T(n)$  for

$n_0 \leq n < 2n_0$  can be considered as constants, we can get the max of them

and  $c_1$ . Now we do similar analysis for lower bound

**Arrays are passed by copying only the subrange:**

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

Let's use the master theorem  $a = 1, b = 2, n^{\log_b a} = 1, f(n) = c_2 n \rightarrow$

Does exist  $e > 0$  such that  $c_2 n = O(n^{-e})$ ? Obviously no, since this would imply that  $c_2 n^{1+e}$  would be bounded above. Which clearly isn't.

Does exist  $k > 0$  such that  $c_2 n = \Theta(\log^k n)$ ? No This would imply that

$$\lim_{n \rightarrow \infty} \frac{c_2 n}{\log^k n} = \infty, \text{ but this limit is } 0$$

Case 3, does exist  $E > 0$  such that  $\Theta(n) = \Omega(n^E)$ . Yes take  $E = 1$

Now let's verify the regularity  $\Theta\left(\frac{n}{2}\right) \leq c_3 n$  for  $c_3 < 1$ ? Not necessarily

Let's guess it's  $\Theta(n)$ . our work in each node is  $n, \frac{n}{2}, \frac{n}{4} \dots$

Our height is such that  $\frac{n}{2^{\text{height}-1}} = 1 \rightarrow \text{height} = \log_2 n + 1 \rightarrow$

$$S = \sum_{k=0}^{\log_2 n} \frac{n}{2^k} \rightarrow \frac{1}{2}S = \sum_{k=0}^{\log_2 n} \frac{n}{2^{k+1}} \rightarrow \frac{1}{2}S = n - \frac{n}{2 \cdot 2^{\log_2 n}} = n - \frac{1}{2} = 2n - 1$$

We will suppose that  $T(n) = \Theta(n) \rightarrow$  So suppose that  $T(k) \leq c_1 k$  for  $n_0 \leq k < n$

$$T(n) = T\left(\frac{n}{2}\right) + n \leq c_1 \frac{n}{2} + n \rightarrow n \left(\frac{c_1}{2} + 1\right). \text{ Take } c_1 \frac{c_1}{2} + 1 = c_2 \rightarrow c_2 = 2. \text{ Take } n_0 = 2$$

Now we need to analyse the case where  $n_0 \leq n < 2n_0 \rightarrow n = 2$  and  $n = 3$ . In this cases

$T(2) = T(3) = 1 \leq 2 * 2 \leq 2 * 3$ . So it's proved that is  $O(n)$ . We argue similarly for the lower bound situation

**Merge Sort:**

**Arrays are passed by a pointer:**

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

$a = 1, b = 2, \log_b a = 1 \rightarrow \Theta(n) = O(n^{1-E})$  Obviously no, this would imply that  $n^{-E}$  is bounded above, which clearly isn't. Case 2. There exist  $k > 0$  such that

$$\Theta(n) = \Theta(n \log^k(n)) \text{? Yes take } k = 0 \rightarrow T(n) = \Theta(n \log(n))$$

**Arrays are passed by copying:**

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

Lets analyzethe work, We know the the height =  $\log_2 n + 1$ , the work done is  $2\Theta(N) + c_1 n, 4\Theta(N) + c_1 \frac{n}{2}, 8\Theta(N) + c_1 \frac{n}{4} \dots$

$$\text{We will guess then that } S = \sum_{k=1}^{\log_2 n+1} 2^k \Theta(N) + \sum_{k=1}^{\log_2 n+1} c_1 \frac{n}{2^{k-1}} =$$

$$S = \Theta(N) \sum_{k=1}^{\log_2 n+1} 2^k + c_1 \sum_{k=0}^{\log_2 n} \frac{n}{2^k}$$

$$\frac{1}{2}S = \Theta(N) \sum_{k=1}^{\log_2 n+1} 2^{k-1} \Theta(N) + c_1 \sum_{k=0}^{\log_2 n} \frac{n}{2^{k+1}} \rightarrow \frac{1}{2}S$$

$$= \Theta(N) \left( \sum_{k=1}^{\log_2 n+1} 2^k - 2^{k-1} \right) + c_1 \left( \sum_{k=0}^{\log_2 n} \frac{n}{2^k} - \frac{n}{2^{k+1}} \right) \rightarrow$$

$$\frac{S}{2} = \Theta(N)(2^{\log_2 n+1} - 1) + c_1 n \left( 1 - \frac{1}{2^{\log_2 n+1}} \right) \rightarrow S = \Theta(N)(2n - 1) + c_1 n - \frac{c_1}{2}$$

We will suppose that it's  $N^2$ , when beginning,  $n = N$

Suppose  $T(k) \leq c_1 n^2 - c_4 N - c_5 n$  for  $n_0 \leq k < n$ . Then

$$T(n) = 2T\left(\frac{n}{2}\right) + c_2 N + c_3 n \leq \frac{c_1 n^2}{2} - 2c_4 N - c_5 n + c_2 N + c_3 n \leq$$

$$c_1 n^2 - 2c_4 N - c_5 n + c_2 N + c_3 n = c_1 n^2 - (2c_4 - c_2)N - (c_5 - c_3)n.$$

$$-(2c_4 - c_2)N + c_3 n \leq -c_4 N \rightarrow -2c_4 N + Nc_2 + c_3 n \leq -c_4 N \rightarrow$$

$$c_4 N \geq Nc_2 + Nc_3 \geq Nc_2 + nc_3 \rightarrow \text{So if we force } c_4 N \geq N(c_2 + c_3) \text{ we are done.}$$

So get  $c_4 = c_3 + c_2$  and then  $T(n) \leq c_1 n^2 - c_4 N$ . We need to verify the case  $T(n)$  for  $n_0 \leq n < 2n_0$ . We need to choose our  $n_0$  the  $\max n_0$ 's that guarantees  $0 \leq \Theta(N) \leq c_4 N$  and  $0 \leq \Theta(n) \leq c_3 n$

Now we need to verify  $T(n)$  for  $n \leq n < 2n_0$ . Since its a well ordered set, it has a maximun. since  $c_4 N$  is a constant, there exists  $c_1$  that makes  $c_1 n^2$

For some  $c_1$ , now just get the max, between this  $c_1$  value and the max of the set of running times. With this we prove the upper bound to be  $O(n^2)$

We can prove the lower bound of this in a analogous manner.

**Arrays are passed by copying only the subrange**

$$T(N, n) = 2T\left(N, \frac{n}{2}\right) + \Theta(n) + \Theta(n)$$

This is equal to  $T(n) = 2T\left(\frac{n}{2}\right) + \Theta(n)$  what is the typical mergesort strategy, by the master theorem we can verify that is  $\Theta(n \lg(n))$ ;

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### 4.3 Solving recurrences with a change of variable

Sometimes a little algebraic manipulation can make an unknown recurrence similar to one you have seen before. Let's solve the recurrence

$$T(n) = 2T(\sqrt{n}) + \Theta(\lg(n))$$

a) Define  $m = \lg(n)$  and  $S(m) = T(2^m)$ .

Rewrite recurrence above in terms of  $m$  and  $S(m)$

$$S(m) = T(2^m) = 2T(\sqrt{2^m}) + \Theta(\lg(2^m)) = S(m) = T(2^m) = 2T(\sqrt{2^m}) + \Theta(m)$$

$$S(m) = 2S\left(\frac{m}{2}\right) + \Theta(m)$$

b) Solve your recurrence for  $S(m)$

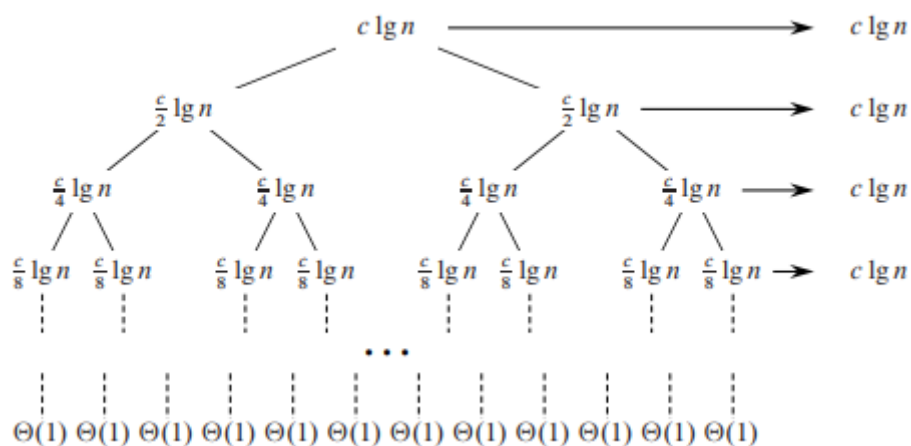
This recurrence is clearly  $m \lg(m)$

c) Use your recurrence for  $S(m)$  to conclude that  $T(n) = \Theta(\lg(n) \lg(\lg(n)))$

$$S(m) = T(2^m) = \lg(n) \lg(\lg(n)) = T(n)$$

d) Sketch the recursion tree for recurrence above and use it to explain intuitively why The solution is  $T(n) = \Theta(\lg(n) \lg(\lg(n)))$

The recursion tree is shown below



We will do  $c \lg(n)$  in each level and have a  $\lg(\lg(n))$  as height, Why?

The size of the problems are  $n, n^{\frac{1}{2}}, n^{\frac{1}{4}}, \dots$

So we have  $n^{\frac{1}{2^{\text{height}-1}}} = 2$  (We don't use 1 because we will never attain it, but we can Guarantee that 2 is the base case)  $\log(\log(n)) + 1 = \text{height}$

$$e) T(n) = 2T(\sqrt{n}) + \Theta(1)$$

$$\text{Let } n = 2^m \text{ and } S(m) = T(2^m) \rightarrow S(m) = 2T(\sqrt{2^m}) + \Theta(1) = S(m) = 2S\left(\frac{m}{2}\right) + \Theta(1)$$

Lets use case 2 of the master theorem Then exist  $k \geq 0$  such that  $\Theta(1) = \Omega(\lg^k(n))$

This implies that  $S(m) = \Theta(\lg(m)) \rightarrow T(n) = \Theta(\lg(n))$

**f)  $T(n) = 3T(\sqrt[3]{n}) + \Theta(n)$**

We will use an analogous guess, take  $S(m) = T(3^m)$  and  $n = 3^m \rightarrow$

$$S(m) = T(3^m) = 3T\left(\sqrt[3]{3^m}\right) + \Theta(3^m) \rightarrow S(m) = 3S\left(\frac{m}{3}\right) + \Theta(3^m)$$

Lets try to solve this recurrence using the tree method, In the first level we have  $3^m$ , in the second we have  $3 \cdot 3^{\frac{m}{3}}$ , in the third level we have  $3 \cdot 3 \cdot 3^{\frac{m}{9}}$ , we can make an induction to prove the work on each node, but its not necessarily since we will prove by Strong inuction. Now lets guess the height

$\frac{m}{3^{i-1}} = 1 \rightarrow \log_3 m + 1 = \text{height. So the sum of work done will be:}$

$$S = \sum_{k=0}^{\log_3 m} 3^k \cdot 3^{\frac{m}{3^k}} \rightarrow \text{But the characteristic of the sum, it looks that the first term}$$

is the most important so we will guess that  $S(m) = \Theta(3^m)$

Now we need to prove this guess, Suppose that  $S(k) \leq c_1 3^k$  for  $n_0^* \leq k < s \rightarrow$

$$S(m) = 3S\left(\frac{m}{3}\right) + \Theta(3^m) \leq 3c_1 3^{\frac{m}{3}} + c_2 3^m \text{ (We know, since its a } \Theta(3^m) \text{ function that}$$

$\exists c_2$  and  $n_0$  such that  $0 \leq \Theta(3^m) \leq c_2 3^m \forall n \geq n_0$ . we then choose  $n_0^* = n_0 \rightarrow$

$$S(m) \leq 3c_1 \frac{3^m}{3^{\frac{2m}{3}}} + c_2 3^m = c_1 3^m \left( \frac{3}{3^{\frac{2m}{3}}} + \frac{c_2}{c_1} \right) \leq c_1 3^m. \text{ We need that } \left( \frac{3}{3^{\frac{2m}{3}}} + \frac{c_2}{c_1} \right) < 1 \rightarrow$$

$$\frac{c_2}{c_1} < 1 + \frac{3}{3^{\frac{2m}{3}}} \rightarrow c_1 > \frac{c_2}{1 + \frac{3}{3^{\frac{2m}{3}}}} \rightarrow c_1 > \frac{c_2}{\frac{3^{\frac{2m}{3}} + 3}{3^{\frac{2m}{3}}}} \rightarrow c_1 > \frac{3^{\frac{2m}{3}} c_2}{3^{\frac{2m}{3}} + 3}. \text{ Now lets analyze}$$

$$\lim_{m \rightarrow \infty} \frac{3^{\frac{2m}{3}} c_2}{3^{\frac{2m}{3}} + 3} = c_2. \text{ This means that there is a } n_0^{**} \text{ such that if we take } c_1$$

$= 2c_2$  and make our  $n_0 = n_0^{**}$  we are safe and we guarantee

that , as long as  $m \geq 3n_0$   $S(m)$  is valid, if all previous are valid. Now we need to guarantee that  $n_0 \leq m$

$< 3n_0$  holds the situation, again we can use the well odering principle, deal with max's

and choose a proper  $c_1$ , Again do a similarly approach for the lower bound considering min's

We then can prove that  $S(m) = \Theta(3^m) \rightarrow S(m) = T(3^m) = T(n) = \Theta(n)$

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#### 4.4 More recurrences examples

Give asymptotically tight upper and lower bounds for  $T(n)$  in each of the following recurrences. Justify your answers:

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

$a = 5, b = 3$ . Does exist  $e > 0$  such that  $n \lg n = O(n^{\log_3 5 - e})$ ? if exists then:

$$0 \leq n^{1+0.4} \lg(n) \leq c_1 n^{\log_3 5} \leftrightarrow \lim_{n \rightarrow \infty} \frac{\lg(n)}{n^{0.065}} = 0. \text{ So take } e = 0.4 \text{ and the case 1 holds}$$

We then have  $T(n) = \Theta(n^{\log_3 5})$

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

Does exist  $e > 0$  such that  $\frac{n}{\lg(n)} = O(n^{-e})$ ? if exist such an  $E$  than exist  $c_1$  such that:

$0 \leq \frac{n}{\lg(n)} \leq \frac{c_1}{n^e} \rightarrow$  This would imply that  $\frac{n^{1+e}}{\lg(n)} \leq c_1$  but this isn't the case, the limit of the last fraction is  $\infty$ . Lets see case 2, does we have  $k \geq 0$  such that  $\frac{n}{\lg(n)} = \Omega(n \lg^k n)$ .

obviously not. Lets see the case 3. Does exist  $e > 0$  such that  $\frac{n}{\lg(n)} = \Theta(n^{1+E})$ .

No. Lets make a guess about the tree. In our first node we have a work of  $\frac{n}{\lg(n)}$ ,

the second node we have  $\frac{n}{\lg(\frac{n}{3})}$ , in the third we have  $\frac{n}{\lg(\frac{n}{9})}$  ... The height is  $\log_3 n + 1$

$$\sum_{k=0}^{\log_3 n} \frac{n}{\lg\left(\frac{n}{3^k}\right)} = n \sum_{k=0}^{\log_3 n} \frac{1}{\lg\left(\frac{n}{3^k}\right)} = n \log_3 2 \sum_{k=0}^{\log_3 n} \frac{1}{\log_3\left(\frac{n}{3^k}\right)} = n \log_3 2 \sum_{k=0}^{\log_3 n} \frac{1}{\log_3 n - k} =$$

$n \log_3 2 \sum_{k=0}^{\log_3 n} \frac{1}{k}$ , we have a problem for the leaves. We do  $\Theta(n)$  in them, since each one

$$\text{So } S = \Theta(n) + n \log_3 2 \sum_{k=1}^{\log_3 n} \frac{1}{k}.$$

We will guess than that  $T(n) = O\left(n \sum_{k=1}^{\log_3 n} \frac{1}{k}\right) \rightarrow$

$$T(k) \leq c_1 k \sum_{i=1}^{\log_3 k} \frac{1}{i}. \text{ for } n_0 \leq k < n \rightarrow$$

$$T(n) = 3T\left(\frac{n}{3}\right) + \frac{n}{\lg(n)} \leq 3c_1 \cdot \frac{n}{3} \sum_{i=1}^{\log_3 \frac{n}{3}} \frac{1}{i} + \frac{n}{\lg(n)} =$$

$$c_1 n \sum_{i=1}^{\log_3 \frac{n}{3}} \frac{1}{i} + \frac{n}{\lg(n)} + c_1 n \sum_{i=\log_3 \frac{n}{3}+1}^{\log_3 n} \frac{1}{i} - c_1 n \sum_{i=\log_3 \frac{n}{3}+1}^{\log_3 n} \frac{1}{i} =$$



$$c_1 n \sum_{i=1}^{\log_3 n} \frac{1}{i} + \frac{n}{\lg(n)} - c_1 n \sum_{i=\log_3 \frac{n}{3}+1}^{\log_3 n} \frac{1}{i} \leq c_1 n \sum_{i=1}^{\log_3 n} \frac{1}{i} \text{ if } \frac{n}{\lg(n)} - c_1 n \sum_{i=\log_3 \frac{n}{3}+1}^{\log_3 n} \frac{1}{i} < 0$$

$c_1 > \frac{1}{\lg(n) \cdot \sum_{i=\log_3 \frac{n}{3}+1}^{\log_3 n} \frac{1}{i}}$ . The limit on the equation in the right goes to  $P$  So we can

find a suitable  $c_1$  given a valid  $n_0$ , after that we can analyze the values between  $n_0 \leq n < 3n_0$ . We can get the maximum of them and  $c_1$  in a way that guarantees the The strong induction hypothesis. Now we have to do the same for the lower bound

Suppose  $T(k) \geq c_2 n \sum_{k=1}^{\log_3 n} \frac{1}{k}$ . for  $n_0 \leq k < n$

Then we would have  $T(n) = 3T\left(\frac{n}{3}\right) + \frac{n}{\lg(n)} \geq 3c_1 \cdot \frac{n}{3} \sum_{i=1}^{\log_3 \frac{n}{3}} \frac{1}{i} + \frac{n}{\lg(n)} \rightarrow$

$$c_1 n \sum_{i=1}^{\log_3 n} \frac{1}{i} \leq c_1 n \sum_{i=1}^{\log_3 n} \frac{1}{i} + \frac{n}{\lg(n)} - c_1 n \sum_{i=\log_3 \frac{n}{3}+1}^{\log_3 n} \frac{1}{i} \leq T(n)$$

$$+ \frac{n}{\lg(n)} - c_1 n \sum_{i=\log_3 \frac{n}{3}+1}^{\log_3 n} \frac{1}{i} > 0 \rightarrow c_1 \leq \frac{1}{\lg(n) + \sum_{i=\log_3 \frac{n}{3}+1}^{\log_3 n} \frac{1}{i}}. \text{ Again this limit goes to a } P$$

We need again analyze the situation where  $n_0 \leq n$

$< 3n_0$ . We analyze the  $c_1$  with the minimums running time of the values in this range

$$c) T(n) = 8T\left(\frac{n}{2}\right) + n^3 n^{\frac{1}{2}}$$

Case 3, lets analyze, does exist  $e > 0$  such that  $n^{3.5} = \Omega(n^{3+0.5=e})$  the regularity:

$$8\left(\frac{n}{2}\right)^{3.5} < cn^{3.5} \frac{8}{2^{3.5}} < 1. \text{ We can find such a } c < 1. \text{ So } T(n) = \Theta(n^{3.5})$$

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

It looks the merge sort algorithm, maybe the 2 doesn't make much difference

Lets suppose  $T(k) \leq c_1 k \lg k - c_2 k$   $n_0 \leq k < n$

$$T(n) = 2T\left(\frac{n}{2} - 2\right) + \frac{n}{2} \leq 2c_1 \left(\frac{n}{2} - 2\right) \log\left(\frac{n}{2} - 2\right) + \frac{n}{2}:$$

$$(c_1 n - 4c_1) \log\left(\frac{n}{2} - 2\right) + \frac{n}{2} - 2c_2 \left(\frac{n}{2} - 2\right) \leq c_1 n \log(n) + \frac{n}{2} - c_2 n \leq c_1 n \log(n)$$

just take  $c_2 = 1$ . This is valid for  $2n_0 + 4 \leq n$ . So we need to analyze:

$n_0 \leq n < 2n_0 + 4$ . By the well ordering principle we can have a maximum

Based on that we can decide what is the best suitable constant for  $c_1$

Now we need to prove the lower bound:

Suppose  $c_2 k \lg k \leq T(k)$  for  $n_0 \leq k < n$  then we would have:

$$2c_1 \left(\frac{n}{2} - 2\right) \lg \left(\frac{n}{2} - 2\right) + \frac{n}{2} \leq T(n)$$

$$cnlg \left(\frac{n}{4}\right) - 4clg \left(\frac{n}{2}\right) + \frac{n}{2} [if \ n \geq 8] \leq c_1 n \lg \left(\frac{n}{2} - 2\right) - 4c_1 \lg \left(\frac{n}{2} - 2\right) + \frac{n}{2} \leq T(n)$$

$$cnlg(n) - 2cn - 4clgn + 4c + \frac{n}{2} \leq cnlg \left(\frac{n}{4}\right) - 4clg \left(\frac{n}{2}\right) + \frac{n}{2} \leq T(n)$$

$$cnlg(n) \leq T(n) \text{ as long as } -2cn - 4clgn + 4c + \frac{n}{2} \geq 0.$$

$$\frac{n}{(4n - 8 + 8 \lg(n))} \geq c \text{ If we analyze the limit as } n \rightarrow \infty \text{ we see its } 0.25. \text{ So get}$$

$$c = \frac{1}{10} \text{ and this will hold. Nos we have to analyze the cases where } n_0 \leq n < 2n_0 + 4$$

We need to analyze the minimum and choose a proper  $c$  based on them

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

It's analogous to the problem b, but now we are in base 2. the height is  $\log_2 n + 1$

In each level we have the following work done:  $\frac{n}{\lg(n)} \rightarrow \frac{n}{\lg\left(\frac{n}{2}\right)} \rightarrow \frac{n}{\lg\left(\frac{n}{4}\right)} \dots$

So we have the following sum for all levels:

$$\sum_{k=1}^{\log_2 n + 1} \frac{n}{\lg\left(\frac{n}{2^{i-1}}\right)}, \text{ we just have a problem at the last level that has } \Theta(1) \text{ cost and there are } n$$

$$\text{So we adjust the equation for } \Theta(n) + \sum_{k=1}^{\log_2 n} \frac{n}{\lg\left(\frac{n}{2^{i-1}}\right)} = \Theta(n) + \sum_{k=0}^{\log_2 n - 1} \frac{n}{\lg\left(\frac{n}{2^i}\right)} =$$

$$n \sum_{k=0}^{\log_2 n - 1} \frac{1}{\lg(n) - i} = n \sum_{k=0}^{\log_2 n - 1} \frac{1}{i}.$$

From here we have the same analysis as the b) case. The limit of this converges to a constant too

$$f) T(n) = T\left(\frac{n}{2}\right) + T\left(\frac{n}{4}\right) + T\left(\frac{n}{8}\right) + n$$

Lets guess the Complexity of this recurrence. In the first level we do  $n$

$$\text{In the second we do } \frac{n}{2} + \frac{n}{4} + \frac{n}{8}$$

$$= \frac{7n}{8}. \text{ In the third we do } \left(\frac{7}{8}\right)^2 n. \text{ This is not a complete tree}$$

We can have an upper and lower bound for the levels the all left leafs are  $\frac{n}{2^{i-1}}$  What gives us

$\log_2 n + 1$ . and for the lower bound we have  $\log_8 n + 1$

Lets analyze both sums:

$$S = n \sum_{k=0}^{\log_2 n} \left(\frac{7}{8}\right)^k \rightarrow \frac{7S}{8} = n \sum_{k=0}^{\log_2 n} \left(\frac{7}{8}\right)^{k+1} \rightarrow S - \frac{7S}{8} = \frac{S}{8} = \sum_{k=0}^{\log_2 n} \left(\frac{7}{8}\right)^k - \left(\frac{7}{8}\right)^{k+1} =$$

$$S = 8 \cdot \left(1 - \left(\frac{7}{8}\right)^{\log_2 n + 1}\right) n = 8n - \frac{8 \cdot 7 \cdot 7^{\log_2 n}}{8 \cdot (2^3)^{\log_2 n}} = 8n - \frac{7 \cdot n^{2.8}}{n^2} \rightarrow \Theta(n)$$

$$\text{Now lets analyze with the upper bound } S = 8 \cdot \left(1 - \left(\frac{7}{8}\right)^{\log_8 n + 1}\right) n = 8n - 7 \cdot n^{0.93}.$$

Our guess is that  $T(n)$  is  $\Theta(n)$ . Let's make a strong inductive hypothesis

Suppose that  $T(k) \leq c_1 n$  for  $n_0 \leq k < n$  then :

$$T(n) = T\left(\frac{n}{2}\right) + T\left(\frac{n}{4}\right) + T\left(\frac{n}{8}\right) + n \leq \frac{c_1 n}{2} + \frac{c_1 n}{4} + \frac{c_1 n}{8} + n = n\left(\frac{c_1}{2} + \frac{c_1}{4} + \frac{c_1}{8} + 1\right) = nc_1\left(\frac{7}{8} + \frac{1}{c_1}\right)$$

take  $c_1 = 8$  Now we need to verify for  $n_0 \leq n < 8n_0 \rightarrow$

$n_0$ . Again we can take  $c_1 = \max\{T(n), \text{for } n \in [n_0, 8n_0), c_1\}$

In this way we guarantee that  $T(n) \leq c \leq cn$  and that  $nc_1 \leq nc$

We need to validate the lower bound now. Our guess will be the same here, so we have:

$$T(n) = T\left(\frac{n}{2}\right) + T\left(\frac{n}{4}\right) + T\left(\frac{n}{8}\right) + n \geq \frac{c_1 n}{2} + \frac{c_1 n}{4} + \frac{c_1 n}{8} + n = n\left(\frac{c_1}{2} + \frac{c_1}{4} + \frac{c_1}{8} + 1\right) = nc_1\left(\frac{7}{8} + \frac{1}{c_1}\right)$$

take  $c_1 = 8$  Now we need to verify for  $n_0 \leq n < 8n_0 \rightarrow$

$$\text{take } c = \min\left\{\frac{T(n)}{n}, \text{for } n \in [n_0, 8n_0), c_1\right\} \text{ then } \frac{T(n)}{n} \geq c \rightarrow cn \leq T(n) \text{ and } cn \leq c_1 n$$

So we proved that  $T(n)$  is  $\Theta(n)$

$$g) T(n) = T(n-1) + \frac{1}{n}$$

Lets make our tree again, it clearly has  $n$  as height. The work done in each level is  $\frac{1}{n}$

$$\text{So our guess will be } T(k) \leq c_1 \sum_{i=1}^k \frac{1}{i} \text{ for } n_0 \leq k < n$$

$$T(n) = T(n-1) + \frac{1}{n} \leq c_1 \sum_{i=1}^{n-1} \frac{1}{i} + \frac{1}{n}. \text{ Take } c_1 = 1, \text{ then } T(n) \leq c_1 \sum_{i=1}^n \frac{1}{i}$$

For the lower bound we have the same case and the proof is straight forward

And again, we need to check  $n_0 \leq n < n_0 + 1$ . So we just need to check the base case,

That, since its an algorithmic recursion, is  $\Theta(1)$

$$h) T(n) = T(n-1) + \lg(n)$$

Again we have  $n$  as height but  $\log_2 n$  as work. Our guess will be:

$$T(k) \leq c_1 \sum_{i=1}^k \log_2 i = \log_2 \left( \prod_{i=1}^k i \right) \rightarrow T(k) \leq c_1 \log_2 k! \leq c_1 k \log(k)$$

Ok, so suppose that  $T(k) \leq c_1 k \log(k)$  for  $n_0 \leq k < n$ . Now we have:

$$T(n) = T(n-1) + \log(n) \leq c_1(n-1) \log(n-1) + \log n \rightarrow$$

$$c_1 n \log(n-1) - c_1 \log(n-1) + \log(n) \leq c_1 n \log(n-1) + \log(n) \\ \leq c_1 \log(n-1) + \log(n) \leq \log(n(n-1)) \leq \log(n!) \leq n \log(n)$$

Just take  $c_1$ . We again would need to analyze for  $n_0 \leq n$

$< n_0 + 1$ , which is the base case and will clearly hold

Now we need to analyze the lower bound. So suppose that  $c_2 k \log(k) \leq T(k)$

for  $n_0 \leq k < n \rightarrow$

$$T(n) = T(n-1) + \log(n) \geq c_2(n-1) \log(n-1) + \log(n)$$

$$c_2(n-1) \log(n-1) + \log(n) \leq T(n) \rightarrow c_2 n \log(n-1) - c_2 \log(n-1) + \log(n) \leq T(n)$$

Take  $n_0$  such that

$c_2 \log(n) \leq c_2(n-1) \log(n-1) + \log(n) \leq T(n)$ . Take  $c_2 = 0.1$  and  $n_0 \geq 2$   
 Again we need to analyze for the case  $n_0 \leq n < n_0 + 1$  which is just the base case  
 which is equal to 1, so we have  $0.1 \log(2) \leq \Theta(2) = 1$ . So it holds

$$i) T(n) = T(n-2) + \frac{1}{\lg n}$$

Lets analyze the recursion tree of the problem above. We have  $n - 2^{i-1}$   
 for the  $i$ -th level. Then  $n - 2(i-1) = 2 \rightarrow n = 2 \text{height} \rightarrow \text{height} = \frac{n}{2}$

Now the work done on level 1 is  $\frac{1}{\lg(n)}$ , level 2 =  $\frac{1}{\log(n-2)}$ , in level 3  $\rightarrow \frac{1}{\log(n-4)} \rightarrow$

$$\sum_{k=1}^{\frac{n}{2}} \frac{1}{\log(n-2(k-1))} = \sum_{k=0}^{n-2} \frac{1}{\log(2+k)}. \text{ The smallest term is } \frac{1}{\lg(n)} \text{ and there are } \frac{n}{2} \text{ items}$$

We will assume that  $T(n) = \Theta\left(\frac{n}{\lg n}\right)$

So suppose that  $T(k) \leq \frac{c_1 k}{\log(k)}$  for  $n_0 \leq k < n$  then we would have:

$$T(n) = T(n-1) + \frac{1}{\lg(n)} \leq \frac{c_1(n-2)}{\log(n-2)} + \frac{1}{\log(n)} \leq \frac{c_1 n}{\log(n)}. \text{ Take } c_1 = 10 \text{ and } n > n_0 = 5$$

Now we need to analyze for  $n_0 \leq n$

$< 5n_0$ . By the well ordering principle there exist such  $c$   
 the  $n_0$  can remain, since the function on the right will grow fast for  $c_1$ , it will obviously grow  
 faster for a  $c_1$  bigger

Now lets prove the lower bound:

$$\frac{c_2 n}{\log(n)} \leq \frac{c_1(n-2)}{\log(n-2)} + \frac{1}{\log(n)} \leq T(n). \text{ Take } c_2 = 10^{-5} \text{ and it hold for } n > 3$$

Again we find a  $c_2$  lower than what is need to contemplate the cases  
 for  $n_0 \leq n < 3n_0$  and take it. the  $n$  will still hold

$$j) T(n) = \sqrt{n} T(\sqrt{n}) + n$$

we again will use the guess by the recursive tree call method. We wont achieve  $T(1)$ .

So we will choose  $T(2)$  as our base case. then we have  $n 2^{\frac{1}{\text{height}-1}} = 2 \rightarrow$

$$n = 2^{2^{\text{height}-1}} \rightarrow \log(n) = 2^{\text{height}-1} \rightarrow \log(\log(n)) = \text{height} - 1 \rightarrow \text{height} = \log(\log(n)) + 1. \text{ Now our summation: is } n \text{ in the first level } n \text{ in the second}$$

at level 3 we have  $\frac{1}{n^4} \frac{1}{n^2} = \frac{3}{n^4}$  nodes, each doing  $\frac{1}{n^4}$  so  $n$ .

$\log(\log(n))+1$

$$\sum_{k=1} n \rightarrow n \log(\log(n))$$

Now we will make our inductive Hypothesis  $T(k) \leq c_1 k \log(\log(k))$  for  $n_0 \leq k < n$ :

$$T(n) = \sqrt{n} T(\sqrt{n}) + n \leq \sqrt{n} c_1 \frac{1}{n^2} \log\left(\frac{1}{2} \log(n)\right) + n =$$

$$c_1 n \log(\log(n)) - c_1 n + n \leq c_1 n \log(\log(n)). \text{ if } n > 2$$

So we just need to verify the cases  $n_0 \leq n < 2n_0 \rightarrow T(2) = 1$  and  $T(3) = 1$

The lower bound is similar

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#### 4.5 Fibonacci number

**This problem develops properties of the Fibonacci numbers, which are defined by recurrence. We will explore the technique of generating functions to solve the Fibonacci recurrence. Define the generating function  $F$  as**

$$F(z) = \sum_{i=0}^{\infty} F_i z^i = 0 + z + z^2 + 2z^3 + 3z^4 + 5z^5 + 8z^6 + \dots$$

**a) Show that  $F(z) = z + z(F(z)) + z^2 F(z) \rightarrow$**

$$\begin{aligned} z + z(F(z)) + z^2 F(z) &= z + \sum_{i=1}^{\infty} F_{i-1} z^i + \sum_{i=2}^{\infty} F_{i-2} z^i \rightarrow z + F_0 z + \sum_{i=2}^{\infty} F_{i-1} z^i + F_{i-2} z^i \rightarrow \\ &= z + F_0 z + \sum_{i=2}^{\infty} z^i (F_{i-1} + F_{i-2}) = F_1 z + F_0 z + \sum_{i=2}^{\infty} F_i z^i = \sum_{i=0}^{\infty} F_i z^i = F(z) \end{aligned}$$

**b) Show that  $F(z) = \frac{z}{1-z-z^2} = \frac{1}{\sqrt{5}} \left( \frac{1}{1-\phi z} - \frac{1}{1-\bar{\phi} z} \right)$**

We showed that  $F(z) = z + z(F(z)) + z^2 F(z) \rightarrow$

$$F(z) - z(F(z)) - z^2 F(z) = z = F(z) = \frac{z}{1-z-z^2}$$

the roots of are the golden ratio as showed in previous exercises

$$\begin{aligned} F(z) &= \frac{z}{(1-\phi z)(1-\bar{\phi} z)} = \frac{A}{(1-\phi z)} - \frac{B}{(1-\bar{\phi} z)} = \frac{A(1-\bar{\phi} z) - B(1-\phi z)}{(1-\phi z)(1-\bar{\phi} z)} = \\ \frac{A-B+B\phi z-A\bar{\phi} z}{(1-\phi z)(1-\bar{\phi} z)} &\rightarrow A \text{ must be } B \rightarrow \frac{zB(\phi-\bar{\phi})}{(1-\phi z)(1-\bar{\phi} z)} = \frac{zB\sqrt{5}}{(1-\phi z)(1-\bar{\phi} z)} \rightarrow \\ B &= \frac{1}{\sqrt{5}} \rightarrow F(z) = \frac{1}{\sqrt{5}} \left( \frac{1}{(1-\phi z)} - \frac{1}{(1-\bar{\phi} z)} \right) \end{aligned}$$

**c) Show that**

$$F(z) = \sum_{i=0}^{\infty} \frac{1}{\sqrt{5}} (\phi^i - \bar{\phi}^i) z^i$$

$$\begin{aligned} F(z) &= \frac{1}{\sqrt{5}} \left( \frac{1}{(1-\phi z)} - \frac{1}{(1-\bar{\phi} z)} \right) = \frac{1}{\sqrt{5}} \sum_{i=0}^{\infty} (\phi z)^i - \sum_{i=0}^{\infty} (\bar{\phi} z)^i \\ \rightarrow F(z) &= \sum_{i=0}^{\infty} \frac{1}{\sqrt{5}} (\phi^i - \bar{\phi}^i) z^i \end{aligned}$$

**d) Use part c to prove that  $F_i = \frac{\phi^i}{\sqrt{5}}$  for  $i > 0$ , rounded to the nearest integer**

$$\begin{aligned} \sum_{i=0}^{\infty} F_i z^i &= \sum_{i=0}^{\infty} \frac{1}{\sqrt{5}} (\phi^i - \bar{\phi}^i) z^i \rightarrow \sum_{i=0}^{\infty} z^i \left( F_i - \frac{1}{\sqrt{5}} (\phi^i - \bar{\phi}^i) \right) = 0 \rightarrow \\ F_i - \frac{1}{\sqrt{5}} (\phi^i - \bar{\phi}^i) &= 0 \rightarrow F_i = \frac{1}{\sqrt{5}} \phi^i - \frac{1}{\sqrt{5}} \bar{\phi}^i \rightarrow \end{aligned}$$

Now we now that  $\bar{\phi}^i = -0.61 \rightarrow |\bar{\phi}|^i < 1 \rightarrow$

$$\left| F_i - \frac{1}{\sqrt{5}} \phi^i \right| = \left| -\frac{1}{\sqrt{5}} \bar{\phi}^i \right| \leq \frac{1}{\sqrt{5}} < \frac{1}{2} \rightarrow \left| F_i - \frac{1}{\sqrt{5}} \phi^i \right| < \frac{1}{2} \rightarrow -\frac{1}{2} < F_i - \frac{1}{\sqrt{5}} \phi^i < \frac{1}{2} \rightarrow$$

So rounding to the nearest integer gives us  $F_i$

**e) Prove that  $F_{i+2} \geq \phi^i$**

By the previous exercise we know that  $F_{i+2} \geq \frac{\phi^{i+2}}{\sqrt{5}} - \frac{1}{2} = \phi^i \left( \frac{\phi^2}{\sqrt{5}} - \frac{1}{2\phi^i} \right) \rightarrow$

$F_{i+2} \geq \phi^i \left( \frac{\phi^2}{\sqrt{5}} - \frac{1}{2\phi^i} \right)$ . Now  $1 \leq \left( \frac{\phi^2}{\sqrt{5}} - \frac{1}{2\phi^i} \right) \leq \frac{\phi^2}{\sqrt{5}}$  for some  $i \rightarrow$

$F_{i+2} \geq \phi^i \left( \frac{\phi^2}{\sqrt{5}} - \frac{1}{2\phi^i} \right) \geq \phi^i$ . if  $i > 3$  the condition above is valid; if  $i = 0, 1, 2 \rightarrow$

$F_2 = 1 \geq 1$ ,  $F_3 = 2 > 1.61$ ,  $F_4 = 3 > \phi^2$  So its proved

#### 4.6 Chip testing

**Professor Diogenes has  $n$  supposedly identical integrated – circuit that in principle are capable of testing each other.**

**The professor's test jig accommodates two chips at a time.**

**When the jig is loaded, each chip tests the other and reports wheter it is good or bad ,but the professor cannot trust the answer of a bad chip. Thus, the four possible outcomes of a test are as follows:**

Chip A says	Chip B says	Conclusion
B is good	A is good	both are good, or both are bad
B is good	A is bad	at least one is bad
B is bad	A is good	at least one is bad
B is bad	A is bad	at least one is bad

**a) Show that if at least  $\frac{n}{2}$  chips are bad, the professor cannot necessarily determine which chips are good using any strategy based on this kind of pairwise test Assume that the bad chips can conspire to fool the professor**

Suppose that there are  $\frac{n}{2}$  goods and  $\frac{n}{2}$  bads.

Each good chip will say that there is  $\frac{n}{2} - 1$  good chips and  $\frac{n}{2}$  bad chips.

the bad chip although can say them same. Being totally

Symetric, in this case there is no way to determine who is good and who is bad

**b) Show that  $\left\lfloor \frac{n}{2} \right\rfloor$  pairwise tests are sufficient to reduce the problem to one of nealy half the size. That is, show how to use  $\left\lfloor \frac{n}{2} \right\rfloor$  pairwise tests to obtain a set with at most  $\left\lfloor \frac{n}{2} \right\rfloor$  chips that still has the property that more than half of the chips are good**

Note that since we have more good chips than bad, if we take out one good and one bad chips this property remains the same guarantee that  $n$  is even so take one aside if needed

Now we will deal with  $\left\lfloor \frac{n}{2} \right\rfloor$  pairwise comparisons choosing 2 chips at a time

If they don't say that both are good, them throw them away or you take 1 good and 1 bad

What will make remain the property. Or you take out 2 bads, what still remain the property  
 Now if both says the other is good, or both is good or both is bad  
 In this situation we discard only one chip, maybe you will discard a good chip. But since  
 There is a pair with two goods, in this case maybe you remain with the  
 same number of goods and bads, but in this situation the worst that can happen is that  
 in the remaining pairs to be analyzed there are more bads than goods. But in this situation  
 We necessarily have a pair formed by two bads, this will make us discard a bad and rebalance  
 the good we discarded early

**c) Show how to apply the solution to part b recursively to identify one good chip.  
 Give and solve the recurrence that describes the number of tests needed to identify  
 one good chip.**

By the technique we described in the previous question, we can make  $\frac{n}{2}$  pairwise comparisons  
 reducing the size of the set to at most  $\left\lfloor \frac{n}{2} \right\rfloor$  through  $\left\lfloor \frac{n}{2} \right\rfloor$  comparisons. so we have

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

We will ignore the floors and ceilings, consider  $n$  to be  $2^k$ . So, lets apply the master theorem case 3  
 there exists  $e > 0$  such that  $\frac{n}{2} = \Omega(n^e)$ ? Yes take  $e = 1$

Now lets analyze the regularity test:  $\frac{n}{4} \leq \frac{c_1 n}{2}$ , yes take  $c_1 = \frac{1}{8}$ . So we have that

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

**d) Show how to identify all the good chips with an additional  $\Theta(n)$  pairwise tests**  
 Since you know a good chip, it will always say the truth, so just iterate over all other chips

**4.7 Monge arrays an  $m \times n$  array  $A$  of real numbers is a Monge Array if for all,  $i, j, k$   
 and  $l$ , such that  $1 \leq i < k \leq m$  and  $1 \leq j < l \leq n$ :**

$$A[i, j] + A[k, l] \leq A[i, l] + A[k, j]$$

**In other words, whenever we pick two rows and two column of a Monge array and  
 consider the four elements at the intersection of the rows and the column, the sum of  
 the upper left and lower right elements is less than or equal to the sum of the lower left  
 and upper right elements**

**a) Prove that an array is Monge if and only if for all  
 $i = 1, 2 \dots m - 1$  and  $j = 1, 2 \dots n - 1$  we have**

$$A[i, j] + A[i + 1, j + 1] \leq A[i, j + 1] + A[i + 1, j]$$

if the matrix is Monge then its true, as shown belo:

let  $i + 1 = k$  and  $j + 1 = l \rightarrow$

$$A[i, j] + A[k, l] \leq A[i, l] + A[k, j] \text{ and we have } 1 \leq i < k \leq m \text{ and } 1 \leq j < l \leq n$$

Now lets make the other way. if

$$A[i, j] + A[k, l] \leq A[i, l] + A[k, j] \text{ and we have } 1 \leq i < k \leq m \text{ and } 1 \leq j < l \leq n \text{ then}$$

$$1 \leq i < k \leq m \text{ and } 1 \leq j < l \leq n: \rightarrow A[i, j] + A[k, l] \leq A[i, l] + A[k, j]$$

We know that  $A[k, j] + A[k + 1, j + 1] \leq A[k, j + 1] + A[k + 1, j] \rightarrow$   
 take  $k = i + n$ . Suppose holds for  $n$ :

$A[i, j] + A[k, j + 1] \leq A[i, j + 1] + A[k, j]$   
 Then we have  $A[i, j] + A[k, j + 1] + A[k, j] + A[k + 1, j + 1]$   
 $\leq A[k, j + 1] + A[k + 1, j] + A[i, j + 1] + A[k, j] \rightarrow$   
 $A[i, j] + A[k + 1, j + 1] \leq A[k + 1, j] + A[i, j + 1] \rightarrow$   
 $A[i, j] + A[i + n + 1, j + 1] \leq A[i + n + 1, j] + A[i, j + 1]$   
 $\rightarrow$  So it implies it is a Monge array

**b) The following array is not Monge. Change one element in order to make it a Monge**

We know that if  $A[i, j] + A[i + 1, j + 1]$   
 $\leq A[i, j + 1] + A[i + 1, j]$ , holds then its monge  
 By inspection on each element we find that  
 $23 + 7 \leq 22 + 6$ ? No. but if we change 22 to 24 it holds for every other pair

**c) Let  $f(i)$  be the index of the column containing the leftmost minimum element of row  $i$ . Prove that  $f(1) \leq f(2) \leq \dots \leq f(m)$  for any  $m \times n$  Monge array**

For a monge Array we have:  
 $A[i, j] + A[k, l] \leq A[i, l] + A[k, j]$   
 Now suppose that  $a_j$  and row  $b_k$  are leftmost minimal element of row  
 $a$  and  $b$  and  $j > k$  (they represent the columns) and  $a$   
 $< b$  (In other words, suppose that the assumption we want to prove is false)  $\rightarrow$   
 $A[a, k] + A[b, j] \leq A[a, j] + A[b, k] \rightarrow$   
 $A[a, j] \leq A[a, k]$ , since its the minimum  
 $A[b, k] \leq A[b, j]$ , since its the minimum  
 $A[a, k] + A[b, j] \geq A[a, j] + A[b, k]$ , which implies that  
 $A[a, k] + A[b, j] = A[a, j] + A[b, k]$ . which implies that

case : 1  
 if  $A[a, k] = A[a, j]$  we would have  $A[a, k]$  as the minimum, not  $j$   
 The same case if  $A[b, k] = A[b, j]$   
 case 2:  
 Now if  $A[a, k] > A[a, j] \rightarrow A[b, j] < A[b, k]$  Then  $k$  would not be the minimum for row  $b$   
 case 3:  
 $A[b, k] < A[b, j] \rightarrow A[a, k] < A[a, j]$ . So  $j$  would not be minimal for row  $a$

**d) Here is a description of a divide and conquer algorithm that computes the left most minimum element in each row of an  $M \times n$  Monge Array  $A$ :**

**Construct a submatrix  $A'$  of  $A$  consisting of the even – numbered rows of  $A$ .**  
**Recursively determine the leftmost minimum for each row of  $A'$ . Then compute the leftmost minimum in the odd – numbered rows of  $A$**   
**Explain how to compute the leftmost minimum in the odd numbered rows of  $A$  ( that the leftmost minimum of the even numbered rows is known) in  $\Theta(n + m)$**

By the previous exercise we know that  $f(2i) \leq f(2i + 1) \leq f(2(i + 1)) \rightarrow$   
 So we just need to iterate over  $f(2i)$  and  $f(2(i + 1))$ . which leads us to  
 $f(2(i + 1)) - f(2i) + 1$  tries to find  $f(2i + 1)$  We need to sum it for all



odd numbered rows So 
$$\sum_{k=0}^{\frac{m}{2}-1} f(2(i+1)) - f(2i) + 1 \rightarrow \frac{m}{2} + \sum_{k=0}^{\frac{m}{2}-1} f(2(i+1)) - f(2i) \rightarrow$$

$$\sum_{k=0}^{\frac{m}{2}-1} f(2(i+1)) - f(2i) = \sum_{k=1}^{\frac{m}{2}} f(2i) - \sum_{k=0}^{\frac{m}{2}-1} f(2i) = f(m) - f(0) + \sum_{k=1}^{\frac{m}{2}-1} f(2i) - f(2i) + \frac{m}{2}$$

So we have That We need to iterate over  $f(m) - f(0) + \frac{m}{2}$ . The maximum value  $f(m) - f(0)$  can have is  $n \rightarrow O(n + m)$

**e) Write the recurrence for the running time of the algorithm in part d. Show that its solution is  $O(m + n \lg(n))$ .**

So the idea behind the algorithm is as follows:

I have a number the rows 1,2,3,4,5 ... m We will compute the left most minimum element

By calling a recursive call on the submatrix 1,2,4,6,8...m (if m is even) and the work level step

Is the calculation we have done in the letter d. So the calculation proceeds as:

$$T(m, n) = T\left(\frac{m}{2}, n\right) + \Theta(n + m). \text{ This is independent of } n, \text{ since } n \text{ won't change:}$$

$$T(m) = T\left(\frac{m}{2}\right) + \Theta(n + m) \rightarrow \text{It looks that its } O(m + n \log(m)) \rightarrow$$

Suppose that  $T(k) \leq c_1 k + c_1 n \log(k)$  for  $n_0 \leq k < n \rightarrow$

$$T(m) = T\left(\frac{m}{2}\right) + \Theta(n + m) \leq c_1 n \log\left(\frac{m}{2}\right) + c_1 \frac{m}{2} + c_2 m + c_2 n =$$

$$c_1 n \log(m) - c_1 n + \frac{c_1 m}{2} + c_2 m + c_2 n. \text{ Take } c_1 = 2c_2$$

$$\begin{aligned} &\rightarrow c_1 n \log(m) - 2c_2 n + c_2 n + c_2 m + c_2 m \leq c_1 n \log(m) + c_2 m + c_2 m \\ &= c_1 n \log(m) + c_1(m) \end{aligned}$$

This guarantees only for  $n \geq 2n_0$ . we need to verify for  $n_0 \leq n < 2n_0$

We have done it a plenty of times. By the well ordering principle might exist a max

We can analyze the the values of  $T(m, n)$  and  $c_1 m$

+  $c_1 n \log(m)$  for the values commented above