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CS430

Introduction to Algorithms

Lec 5

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Outlines

- **Recursion Tree**
- **Master Theorem and Extended Form**
- **Selection Sort and Bubble Sort**

Ex5. (recurrence relation)

Prove that $T(n)=O(n)$, while

proof:

$$T(n) = T\left(\left\lfloor \frac{n}{2} \right\rfloor\right) + T\left(\left\lceil \frac{n}{2} \right\rceil\right) + 1$$

our goal is to prove: $T(n) \leq cn$

Hypothesis:

$$T\left(\left\lfloor \frac{n}{2} \right\rfloor\right) \leq c \left\lfloor \frac{n}{2} \right\rfloor$$
$$T\left(\left\lceil \frac{n}{2} \right\rceil\right) \leq c \left\lceil \frac{n}{2} \right\rceil$$

then $T(n) \leq c \left\lfloor \frac{n}{2} \right\rfloor + c \left\lceil \frac{n}{2} \right\rceil + 1 = cn + 1$ ❌

Adjust our goal to $T(n) \leq cn - d$

Hypothesis: $T\left(\left\lfloor \frac{n}{2} \right\rfloor\right) \leq c \left\lfloor \frac{n}{2} \right\rfloor - d$

$$T\left(\left\lceil \frac{n}{2} \right\rceil\right) \leq c \left\lceil \frac{n}{2} \right\rceil - d$$

then:

$$T(n) \leq c \left\lfloor \frac{n}{2} \right\rfloor - d + c \left\lceil \frac{n}{2} \right\rceil - d + 1 = cn - 2d + 1 = cn - (2d - 1)$$

when $2d - 1 > d$, $T(n) \leq cn - d$.

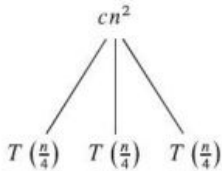
$(2d - 1 > d \Rightarrow d > 1)$, which is easy to satisfy

QED.

Recursion Tree

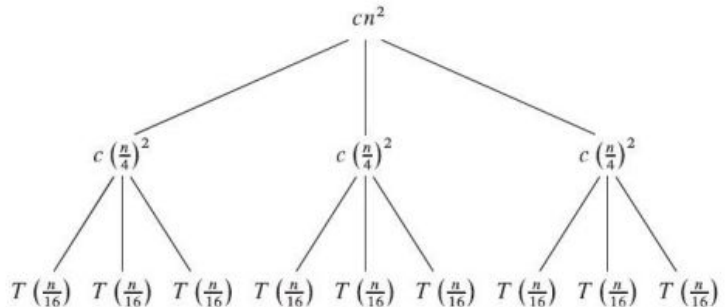
Ex1: Solve the upper bound of $T(n)$, while
 $T(n)=3T(n/4)+cn^2$

$T(n)$

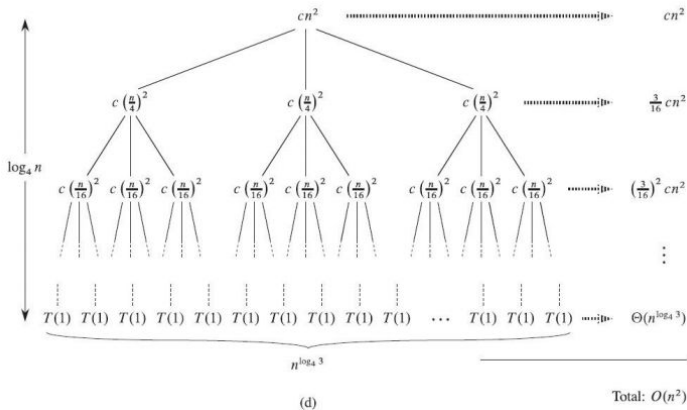


(a)

(b)



(c)



- Number of levels L : $L = \log_4 n$, for more accuracy, $L = \log_4 n + 1$
- Cost of the i^{th} level: $(\frac{3}{16})^{i-1} cn^2$
- How many leaves? $3^{\log_4 n} = n^{\log_4 3}$ Then the cost of all leaves is : $cn^{\log_4 3} = \Theta(n^{\log_4 3})$

Sum up all costs from all levels:

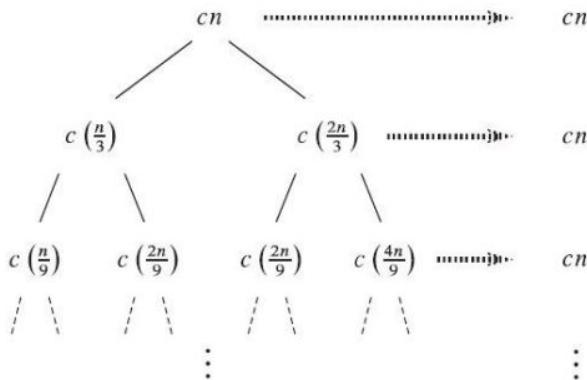
$$T(n) = cn^2 \left[1 + \frac{3}{16} + \left(\frac{3}{16}\right)^2 + \dots + \left(\frac{3}{16}\right)^{\log_4 n} \right] + \Theta(n^{\log_4 3}) < cn^2 \left[1 + \frac{3}{16} + \left(\frac{3}{16}\right)^2 + \dots + \left(\frac{3}{16}\right)^{\infty} \right] + \Theta(n^{\log_4 3})$$

$$\sum_{k=0}^{\infty} x^k = \frac{1}{1-x} \quad T(n) = cn^2 \left(\frac{1}{1 - \frac{3}{16}} \right) + \Theta(n^{\log_4 3}) = \frac{16}{13} cn^2 + \Theta(n^{\log_4 3}) = O(n^2)$$

Last example for Recursion Tree

EX2: Solve the upper bound of

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



The number of levels-- k : $\left(\frac{2}{3}\right)^k n = 1$,

then

$$k = \log_{\frac{3}{2}} n = \frac{\lg n}{\lg \frac{3}{2}} = \frac{\lg n}{\lg 3 - 1} = \frac{1}{\lg 3 - 1} \lg n$$

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

$$T(n) = O(n \lg n)$$

Master Theorem

Applicable for recurrence relation:

$$T(n) = aT(n/b) + f(n)$$

- If your $T(n)$ satisfies any of the following cases, the asymptotic bounds can be solved according to the Master Theorem;
- Not all cases are included in cases of Master Theorem.

Three Cases of Master Theory

Case1:

If $f(n) = O(n^{\log_b a - \varepsilon})$ for some constants $\varepsilon > 0$, then $T(n) = \Theta(n^{\log_b a})$.

Case2:

If $f(n) = \Theta(n^{\log_b a})$ then $T(n) = \Theta(n^{\log_b a} \lg n)$.

Case 3:

If $f(n) = \Omega(n^{\log_b a + \varepsilon})$ for some constant $\varepsilon > 0$ and if $af(n/b) \leq cf(n)$ for some constant $c < 1$ and sufficiently larger n then $T(n) = \Theta(f(n))$.

$$T(n) = aT(n/b) + f(n)$$

$$T(n) = 8T(n/2) + n^2$$

Ex: Solve the upper bound of

$$T(n)=8T(n/2)+n^2$$

Derive Master Theorem, then we have

$$a=8, b=2, f(n)=n^2$$

$$\log_2 8=3, f(n)=O(n^2)=O(n^{3-1}).$$

That is, when $\epsilon=1$, it holds for case 1.

$$\text{So, } T(n)=O(n^3)$$

Case1:

$$T(n)=aT(n/b)+f(n)$$

If $f(n) = O(n^{\log_b a - \epsilon})$ for some constants $\epsilon > 0$, then $T(n) = \Theta(n^{\log_b a})$

Case2:

If $f(n) = \Theta(n^{\log_b a})$ then $T(n) = \Theta(n^{\log_b a} \lg n)$.

Case1:

$$T(n) = aT(n/b) + f(n)$$

If $f(n) = O(n^{\log_b a - \varepsilon})$ for some constants $\varepsilon > 0$, then $T(n) = \Theta(n^{\log_b a})$

Case2:

If $f(n) = \Theta(n^{\log_b a})$ then $T(n) = \Theta(n^{\log_b a} \lg n)$.

Ex: Solve $T(n)$'s asymptotic bound when

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

Solution: $a=1$, $b=3/2$, $f(n)=1$

then $\log_b a = \log_{\frac{3}{2}} 1 = 0$ and $f(n)=1=n^0$

It matches case 2 and gives us: $T(n) = \Theta(\lg n)$

Case 3:

If $f(n) = \Omega(n^{\log_b a + \varepsilon})$ for some constant $\varepsilon > 0$ and if $af(n/b) \leq cf(n)$ for some constant $c < 1$ and sufficiently

large n then $T(n) = \Theta(f(n))$.

$$T(n) = aT(n/b) + f(n)$$

Ex: Solve $T(n)$'s asymptotic bound when

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

Solution: $a=3$, $b=4$, $f(n)=n \lg n$

then $\log_b a = \log_4 3 = 0.8$ and $f(n)=n \lg n$

$$f(n) = \Omega(n) = \Omega(n^{0.8+0.2}) .$$

It matches case 3 when $\varepsilon=0.2$ and gives us:

$$T(n) = \Theta(f(n)) = \Theta(n \lg n)$$

Case1:

If $f(n) = O(n^{\log_b a - \epsilon})$

Case2:

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

Ex: Solve $T(n)$'s asymptotic bound when

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

$a=2, b=2, f(n)=n^2$, then $n^{\log_b a} = n$, that violates the first two cases. (Because the upper bound of n^2 is impossible to be n or $n^{1-\epsilon}$)

Case 3:

If $f(n) = \Omega(n^{\log_b a + \epsilon})$

Let's try case 3.

Prove that $f(n)=n^2=\Omega(n^{1+\epsilon})$.

We can find 1 as the value of ϵ satisfying the above statement.

So it matches case 3 and gives us:

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

Extended Form of Master Theorem

Case 1: if $af(\frac{n}{b}) = cf(n)$ is true for some constant $c < 1$, then $T(n) = \Theta(f(n))$

Case 2: if $af(\frac{n}{b}) = cf(n)$ is true for some constant $c > 1$, then $T(n) = \Theta(n^{\log_b a})$

Case 3: if $af(\frac{n}{b}) = f(n)$ is true, then $T(n) = \Theta(f(n)\log_b n)$.

Proof

$$T(n) = aT\left(\frac{n}{b}\right) + f(n)$$

Prove Extended form of Master Theorem
Recursion Tree



$$f\left(\frac{n}{b}\right) f\left(\frac{n}{b}\right) \dots f\left(\frac{n}{b}\right) \rightarrow a f\left(\frac{n}{b}\right)$$



$$f\left(\frac{n}{b^2}\right) \dots f\left(\frac{n}{b^2}\right) \rightarrow a^2 f\left(\frac{n}{b^2}\right)$$

⋮

⋮



$$f(1) f(1) \dots \rightarrow a^L \cdot f(1) = a^L f\left(\frac{n}{b^L}\right)$$

▲ the number of levels: $L = \log_b n$

▲ the number of leaves: $a^L = a^{\log_b n} = n^{\log_b a}$

▲ Total cost = $f(n) + a f\left(\frac{n}{b}\right) + a^2 f\left(\frac{n}{b^2}\right) + \dots + a^L f\left(\frac{n}{b^L}\right)$
suppose that $a f\left(\frac{n}{b}\right) = c f(n)$, plus it in:
 $a f\left(\frac{n}{b}\right) = c f(n) \leftarrow \rightarrow a^2 f\left(\frac{n}{b^2}\right)$

$$= a \cdot a f\left(\frac{n}{b}\right)$$

$$= a \cdot c f\left(\frac{n}{b}\right)$$

$$= c \cdot a f\left(\frac{n}{b}\right)$$

$$= c \cdot c f(n)$$

$$= c^2 f(n)$$

Then the total cost is:

$$f(n) + c f(n) + c^2 f(n) + \dots + c^L f(n)$$

$$= f(n) (1 + c + c^2 + \dots + c^L)$$

$$= f(n) \frac{c^{L+1} - 1}{c - 1}$$

$$T(n) = f(n) \frac{c^{L+1} - 1}{c - 1} = f(n)(1 + c + c^2 + \dots + c^L)$$

① when $c = 1$, $T(n) = f(n) \cdot (L+1) = f(n)(\log_b n + 1)$
 when $n \rightarrow \infty$, $\Theta(\log_b n + 1) = \Theta(\log_b n)$

$$\therefore T(n) = \Theta(f(n) \cdot \log_b n)$$

② when $c < 1$, $T(n) = \Theta(f(n))$

③ when $c > 1$, last level $= f(n) c^L$
 $c^L = c^{\log_b n} = n^{\log_b c}$ } \Rightarrow
 from the recursion tree: $a^L f(n) = a^L \cdot K$

$f(n) \cdot c^L = f(n) \cdot n^{\log_b c} = a^{\log_b n} \cdot K = n^{\log_b a} \cdot K$
 while $T(n) = \Theta(c^L f(n)) = \Theta(n^{\log_b a} \cdot K)$
 $= \Theta(n^{\log_b a})$

Extended form of Master Theorem

Methodology

Step1: list values of a, b and f(n)

Step2: plug a and b in to evaluate $af(n/b)$

Step3: set $af(\frac{n}{b}) = cf(n)$ and solve the value of c.

Step4: match the value of c to the above 3 cases.

case1: when $c < 1$, $T(n) = O(f(n))$

case 2: when $c > 1$, $T(n) = O(n^{\log_b a})$

case 3: when $c = 1$, $T(n) = O(f(n) \log_b n)$

By Lan Yao

Ex1. Solve the asymptotic bound of $T(n)$, when $T(n) = 3T(\frac{n}{2}) + n$

Handwritten solution on lined paper:

$$a=3, b=2, f(n)=n$$
$$af(\frac{n}{b}) = 3f(\frac{n}{2}) = 3 \cdot \frac{n}{2} = \frac{3}{2}n$$

Suppose: $\frac{3}{2}n = cf(n) = c \cdot n = cn$

then $c = \frac{3}{2} > 1 \rightarrow$ match

Case 2 gives you: $T(n) = \Theta(n^{\log_b a})$
 $= \Theta(n^{\log_2 3})$

case1: when $c < 1$, $T(n) = O(f(n))$
case 2: when $c > 1$, $T(n) = O(n^{\log_b a})$
case 3: when $c = 1$, $T(n) = O(f(n) \log_b n)$

By Lan Yao

Ex2. Solve the asymptotic bound of $T(n)$, when $T(n) = T\left(\frac{3n}{4}\right) + n$

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

$$a=1, \quad b=\frac{4}{3}, \quad f(n)=n$$

$$af\left(\frac{1}{b}\right) = 1 \cdot f\left(\frac{3}{4}n\right) = \frac{3}{4}n$$

$$\text{Suppose that } af\left(\frac{1}{b}\right) = cf(n)$$

$$\text{that is: } \frac{3}{4}n = c \cdot n \Rightarrow c = \frac{3}{4}$$

$$\therefore c = \frac{3}{4} < 1 \quad \therefore \text{case 1}$$

$$\text{Case 1 gives us: } T(n) = \Theta(f(n)) \\ = \Theta(n)$$

case 1: when $c < 1$, $T(n) = \Theta(f(n))$

case 2: when $c > 1$, $T(n) = \Theta(n^{\log_b a})$

case 3: when $c = 1$, $T(n) = \Theta(f(n) \log_b n)$

Ex3. Solve the asymptotic bound of $T(n)$, when $T(n) = 2T(\frac{n}{2}) + n$

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

$$a=2, b=2, f(n)=n$$

$$af(\frac{n}{b}) = 2f(\frac{n}{2}) = 2 \cdot \frac{n}{2} = n$$

$$\text{suppose that } af(\frac{n}{b}) = cn \Rightarrow n = cn$$

$$\Rightarrow c=1 \Rightarrow \text{case 3}$$

$$\text{Case 3 gives us: } T(n) = \Theta(f(n) \cdot \log_b n)$$

$$= \Theta(n \lg n)$$

case1: when $c < 1$, $T(n) = O(f(n))$

case 2: when $c > 1$, $T(n) = O(n^{\log_b a})$

case 3: when $c = 1$, $T(n) = O(f(n) \log_b n)$

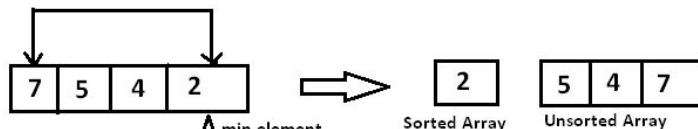
Algorithmic Analysis of other Sorts

- Selection Sort
- Bubble Sort

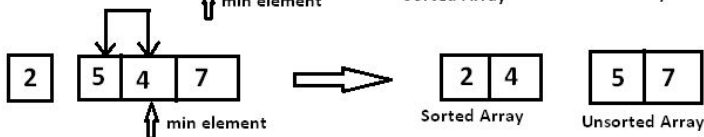
Selection Sort

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STEP 1.



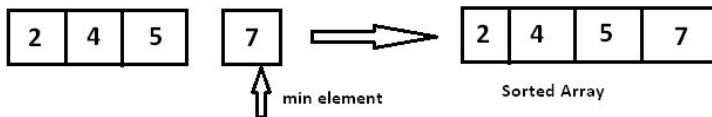
STEP 2.



STEP 3.



STEP 4.



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Complexity of Selection Sort

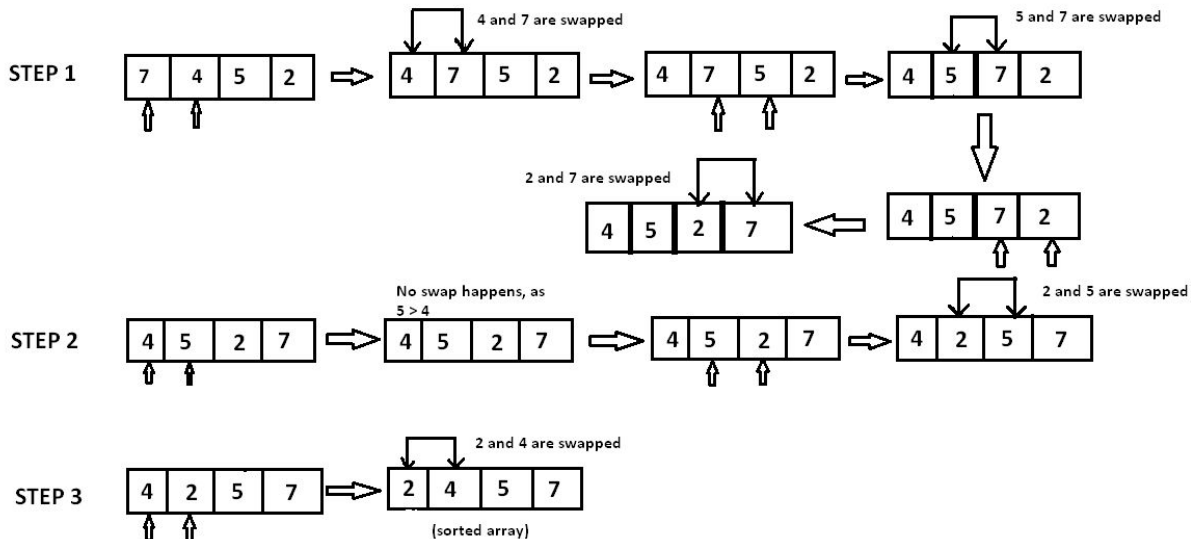
$$\sum_{i=1}^{n-1} i = \frac{n(n-1)}{2} = O(n^2)$$

- Best case: $O(n^2)$
- Worst case: $O(n^2)$
- Average case: $O(n^2)$

Bubble Sort

Repeatedly stepping through the array, comparing adjacent elements and swapping them if they are in a wrong order until the list is sorted, which is confirmed by no swap.

Bubble Sort



Complexity of Bubble Sort

$$\sum_{i=1}^{n-1} i = \frac{n(n-1)}{2} = O(n^2)$$

● Best case: $T(n)=n-1=O(n)$

● Worst case: $T(n)=O(n^2)$

● Average case:

$$\begin{aligned} T(n) &= \sum_{i=1}^{n-1} X_i p_i = \frac{1}{2} \sum_{i=1}^{n-1} X_i = \frac{1}{2} \times \frac{n(n-1)}{2} = \frac{1}{4} n^2 - \frac{1}{4} n \\ &= O(n^2) \end{aligned}$$