

CSE 321 - INTRODUCTION TO ALGORITHMS

Homework #3

1) I used Master's Theorem for some questions. This theorem is:

$$T(n) = a \cdot T\left(\frac{n}{b}\right) + n^c$$

where $c < \log_b a$ then $T(n) = \Theta(n^{\log_b a})$

where $c = \log_b a$ then $T(n) = \Theta(n^c \log n)$

where $c > \log_b a$ then $T(n) = \Theta(f(n)) = \Theta(n^c)$

a) $T(n) = 27T(n/3) + n^2$

$a=27, b=3, c=2$

$\log_b a = \log_3 27 = 3$ So $c < \log_b a$, then $T(n) = \Theta(n^3)$

b) $T(n) = 9T(n/4) + n$

$a=9, b=4, c=1$

$\log_b a = \log_4 9 = 1,58$ So $c < \log_b a$, then $T(n) = \Theta(n^{\log_4 9}) = \Theta(n^{1,58})$

c) $T(n) = 2T(n/4) + n^{\frac{1}{2}}$

$a=2, b=4, c=1/2$

$\log_b a = \log_4 2 = 1/2$ So $c = \log_b a$, then $T(n) = \Theta(\sqrt{n} \cdot \log n)$

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

I will change the formula to simulate Master's Theorem formula.

Let's suppose $m = \log n, n = 2^m$

Now $T(2^m) = 2T(2^{m/2}) + 1$, let's say $S(m) = T(2^m)$ take $\log 2$

$S(m) = 2S(m/2) + 1$ According to Master's Theorem:

$a=2, b=2, c=0$; $\log_b a = \log_2 2 = 1$ So $c < \log_b a$ then $S(m) = \Theta(m)$

We said that $S(m) = T(2^m)$ so $T(2^m) = \Theta(m)$

$T(2^m) = \Theta(m)$ change with 2^m to n

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

e) $T(n) = 2T(n-2)$ $T(0)=1, T(1)=1$

$$\begin{aligned} T(n) &= 2T(n-2) \\ &= 2 \cdot (2T(n-4)) = 2^2 (T(n-2 \cdot 2)) \\ &= 2 \cdot 2 \cdot (2T(n-6)) = 2^3 T(n-2 \cdot 3) \end{aligned}$$

$T(n) = 2^k T(n-2 \cdot k)$, We know that $T(1)=1$

So $n-2 \cdot k = 1 \Rightarrow k = \frac{n-1}{2}$

$T(n) = 2^{\frac{n-1}{2}} \cdot T(1) = 2^{\frac{n}{2}} \times \underbrace{2^{-\frac{1}{2}}}_{\text{constant}}$ So $T(n) = O(2^{\frac{n}{2}})$

f) $T(n) = 4T(n/2) + n$, $T(1)=1$

1st Way

Master's Theorem : $a=4, b=2, c=1$

$\log_b a = \log_2 4 = 2$ So $c < \log_b a$ $T(n) = O(n^2)$

2nd Way

$$\begin{aligned} T(n) &= 4 \cdot T(n/2) + n \\ &= 4 \cdot (4 \cdot T(n/2^2) + n/2) + n \\ &= 4 \cdot 4 \cdot (4 \cdot T(n/2^3) + n/2^2) + n \end{aligned}$$

$T(n) = 4^k T(n/2^k + n/2^{k-1}) + n$, We know that $T(1)=1$

So $n/2^k + 2n/2^k = 1$ $\frac{3n}{2^k} = 1$ $3n = 2^k$ $k = \log_2 3n$

$T(n) = 4^{\log_2 3n} \cdot T(1) + \log_2 3n$
 $2^{\log_2 3n^2} \cdot 1 + \log_2 3n = \underbrace{9n^2}_{\text{big growth rate}} + \log_2 3n$ So $O(\max(n^2, \log_2 3n)) = O(n^2)$

g) $T(n) = 2T(\sqrt{n}) + 1$, $T(3) = 1$

Change formula to simulate Master Theorem's formula

Let's suppose $m = \log_3 n$, $n = 3^m$

$$T(3^m) = 2T(3^{m/2}) + 1 \quad , \quad \text{let's say } S(m) = T(3^m) \quad \text{take } \log 3$$

$$S(m) = T(3^m)$$

$$S(m) = 2 \cdot S(m/2) + 1 \quad \text{According to Master's Theorem:}$$

$$a=2 \quad b=2 \quad c=0$$

$$\log_b a = \log_2 2 = 1 \quad \text{So } c < \log_b a \quad \text{then } S(m) = \Theta(m^{0,63})$$

$$\text{We said that } S(m) = T(3^m) \quad \text{so } T(3^m) = \Theta(m^{0,63})$$

$$T(3^m) = \Theta(m^{0,63}) \quad \text{Change } 3^m \text{ with } n$$

$$T(n) = \Theta(\log_3 n)^{0,63} = \Theta(\log_3 n)^{\log_3 2}$$

2) I wrote this relation

$f(n)$

if $n \leq 1$
print } 1 constant

else

for $i=1$ to $n \rightarrow n$ times

$f(n/2) \rightarrow T(n/2)$

$$T(n) = \begin{matrix} 1 \\ n \times T(n/2) \end{matrix}, n \leq 1, n > 1$$

$$T(1) = 1$$

We assume n is a power of 2

So $n = 2^k, k = \log n$

$$T(2^k) = 2^k \times T(2^{k-1}) + 1$$

$$= 2^k \times (2^{k-1} \times T(2^{k-2}))$$

$$= 2^k \times 2^{k-1} \times (2^{k-2} \times T(2^{k-3}))$$


$$\rightarrow 2^k \times 2^{k-1} \times \dots \times 2 = 2^{\frac{k(k+1)}{2}}$$

So

$$T(2^k) = 2^{\frac{k(k+1)}{2}} \times T(1)$$

$$T(2^k) = (2^k)^{k/2} \times (2^k)^{1/2} \times T(1)$$

$$T(n) = n^{k/2} \times n^{1/2} \Rightarrow n^{\frac{k+1}{2}} \Rightarrow n^{\frac{\log n + 1}{2}}$$

$$T(n) = O(n^{\frac{\log n + 1}{2}}) = O(n^{\log n})$$

For line numbers, I wrote the function with Python format (in file)

Counter examples

<u>n</u>	<u>Line</u>
$2 = 2^1$	$2 = 2^1$
$4 = 2^2$	$8 = 2^2 \cdot 2^1 = 2^3$
$8 = 2^3$	$64 = 2^3 \cdot 2^2 \cdot 2^1 = 2^6$
$16 = 2^4$	$1024 = 2^4 \cdot 2^3 \cdot 2^2 \cdot 2^1 = 2^{10}$
\vdots	\vdots
$n = 2^k$	$= 2^k \cdot 2^{k-1} \cdot 2^{k-2} \dots 2^1 = 2^{\frac{k \cdot (k+1)}{2}}$

For n , there is $2^{\frac{k \cdot (k+1)}{2}}$ line.

$k = \log n$, so for n , $2^{\frac{\log n (\log n + 1)}{2}}$ line printed.

3- That algorithm performs constant time with computation and then recursively calls itself three times. Each time on array whose size $2/3$ of the original size

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

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

$$= 1 + 3 + 9T\left(\frac{4n}{9}\right)$$

$$\vdots$$

$$= 1 + 3 + 3^2 + \dots + 3^{\log_{3/2} n}$$

$$\sum_{k=0}^n a^k = \frac{a^{n+1} - 1}{a - 1} \quad \text{for } a > 1$$

$$\text{So } \frac{3^{\log_{3/2} n + 1} - 1}{3 - 1} \Rightarrow O(3^{\log_{3/2} n})$$

$$= O(3^{(\log_3 n) / (\log_3 3/2)}) \quad \text{change base}$$

$$= O(n^{1 / \log_3 3/2}) \quad \log_3 3/2 = 0,36$$

$$= O(n^{1/0,36}) = O(n^{2,77})$$

4- Average Case Analysis for Quick Sort:

- Assume that it is equally likely that the pivot element $L[low]$ will be fixed in any position after rearrange

$$T = T_1 + T_2 \rightarrow \text{Random Variables}$$

\downarrow \downarrow
 # of operations # of operations
 in rearrange in recursive

$$A(n) = E[T] = E[T_1] + E[T_2] \quad \begin{array}{l} \text{depends on where the} \\ \text{pivot has been placed} \end{array}$$

\hookrightarrow high-low+2 operations (fixed)

$$E[T_2] = \sum_x E[T_2 | \mathcal{X} = x] \cdot P(\mathcal{X} = x)$$

\downarrow \downarrow $\hookrightarrow 1/n$
 position of pivot cond. prob.

$$A(n) = \underbrace{(n+1)}_{E[T_1]} + \sum_{i=1}^n E[T_2 | \mathcal{X} = i] \cdot \underbrace{P(\mathcal{X} = i)}_{= \frac{1}{n}}$$

$$= (n+1) + \sum_{i=1}^n [A(i-1) + A(n-i)] \cdot \frac{1}{n}$$

$A(0) + A(n-1)$
 $A(1) + A(n-2)$
 \vdots
 $A(n-1) + A(0)$
 $\hline 2(A(0) + A(n-1))$

$$\begin{aligned} n \cdot A(n) &= n \cdot (n+1) + 2[A(0) + \dots + A(n-1)] \\ - (n-1)A(n-1) &= n \cdot (n-1) + 2[A(0) + \dots + A(n-2)] \end{aligned}$$

$$nA(n) - (n-1)A(n-1) = 2n + 2A(n-1)$$

$$\frac{1}{n \cdot (n+1)} \text{ multiply by } \Rightarrow \frac{A(n)}{n+1} - \frac{A(n-1)}{n} = \frac{2}{n+1} \quad \left(t(n) = \frac{A(n)}{n+1} \right)$$

$$t(n) = t(n-1) + \frac{2}{n+1}$$

$$A(1) = 0 // \text{initial}$$

$$t(0) = \frac{A(0)}{1} \rightarrow t(n) = \sum_{i=2}^n \frac{2}{i+1} = 2 \underbrace{H(n+1)}_{\text{Harmonic}} - 3$$

$$A(n) = t(n) \cdot (n+1) = 2 \cdot (n+1) \cdot \underbrace{H(n+1)}_{\ln(n+1)} - 3 \cdot (n+1) \in O(n \log n)$$

- Average Case Analysis for Insertion Sort:

Let T_i = # of basic operator at step i ; $1 \leq i \leq n-1$

$$T = T_1 + T_2 + \dots + T_{n-1} = \sum_{i=1}^{n-1} T_i$$

$$A(n) = E[T] = E\left[\sum_{i=1}^{n-1} T_i\right] = E[T_1] + E[T_2] + \dots + E[T_{n-1}]$$

$$E[T_i] = \sum_{j=1}^i j \cdot P(T_i = j)$$

1 comparison will occur if $x = L[i] > L[i-1]$
 2 " " " " $L[i-2] < x < L[i-1]$
 \vdots
 i " " " " $x < L[i]$

There are $(i+1)$ intervals that x can fall in

$$P(T_i = j) = \begin{cases} \frac{1}{i+1} & \text{if } 1 \leq j \leq i-1 \\ \frac{2}{i+1} & \text{if } j = i \end{cases}$$

$$E[T_i] = \left[\sum_{j=1}^{i-1} \left(j \cdot \frac{1}{i+1} \right) \right] + i \cdot \frac{2}{i+1} = \frac{i(i-1)}{2(i+1)} + \frac{2}{(i+1)} = \frac{i^2 - i + 4i}{2(i+1)} = \frac{i(i+3)}{2(i+1)}$$

$$A(n) = E[T] = \sum_{i=1}^{n-1} E[T_i] = \sum_{i=1}^{n-1} \left(\frac{i}{2} + 1 - \frac{1}{i+1} \right) = \frac{n \cdot (n-1)}{4} + n-1 - \underbrace{\sum_{i=1}^{n-1} \frac{1}{i+1}}_{\text{Harmonic}}$$

$$= \frac{n \cdot (n-1)}{4} + n - H(n)$$

$$A(n) \in O(n^2)$$

Now, according to analyzes
QuickSort is $O(n \log n)$
InsertionSort is $O(n^2)$

There are 10 swap count comparisons in my report file.
As a result, I decide the quick sort is faster than insertion sort.

5- a) $T(n) = 5T(n/3) + n^2$

$$a=5, b=3, c=2$$

$$\log_b a = \log_3 5 = 1.46$$

$$\text{So } c > \log_b a \text{ then } T(n) = O(f(n)) = O(n^2)$$

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

$$a=2, b=2, c=2$$

$$\log_b a = \log_2 2 = 1$$

$$\text{So } c > \log_b a \text{ then } T(n) = O(f(n)) = O(n^2)$$

c) $T(n) = T(n-1) + n$

$$= [T(n-2) + n-1] + n$$

$$= T(n-3) + (n-2) + (n-1) + n$$

$$T(n) = T(n-k) + (n-(k-1)) + (n-(k-2)) + \dots + (n-1) + n$$
$$n-k=0 \quad k=n$$

$$T(n) = T(n-n) + (n-n+1) + (n-n+2) + \dots + (n-1) + n$$

$$T(n) = 0 + 1 + 2 + 3 + \dots + n$$

$$T(n) = \frac{n \cdot (n+1)}{2} = \frac{n^2 + n}{2}$$

$$\text{So } T(n) = O(n^2)$$

→ Now, all algorithms have same growth order. But I would choose algorithm C. Because it needs less space and its component is linear time.