

CSE 321 Homework #3

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1-) Master theorem:

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

If $f(n) \in \mathcal{O}(n^d)$ where $d \geq 0$ then,

$$T(n) \in \begin{cases} \mathcal{O}(n^d), & \text{if } a < b^d & \text{case 1} \\ \mathcal{O}(n^d \cdot \log n), & \text{if } a = b^d \text{ for all } n & \text{case 2} \\ \mathcal{O}(n^{\log_b a}), & \text{if } a > b^d & \text{case 3} \end{cases}$$

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

$f(n) \in \mathcal{O}(n^2)$ so $d=2$

$a=27, b=3$

$27 > 3^2$ so; case 3

$T(n) \in \mathcal{O}(n^{\log_3 27}) = \mathcal{O}(n^{\log_3 27})$

$T(n) \in \mathcal{O}(n^3)$

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

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

$9 > 4^1$ so; case 3

$T(n) \in \mathcal{O}(n^{\log_4 9})$ $\log_4 9 \approx 1.58$

$T(n) \in \mathcal{O}(n^{1.58})$

c) $T(n) = 2 T(n/4) + \sqrt{n}$

$a=2, b=4, d=\frac{1}{2}$

$2 = 4^{\frac{1}{2}}$ so; case 2

$T(n) \in \mathcal{O}(\sqrt{n} \cdot \log n)$

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

$$\text{lets say } n = 2^k$$

$$T(2^k) = 2 T(2^{k/2}) + 1 \quad \text{and} \quad T(2^k) = F(k)$$

$$F(k) = 2 F(k/2) + 1$$

now apply this master theorem

$$a=2, b=2, d=0$$

$$2 > 2^0 \text{ so; case 3}$$

$$F(k) \in O(k^{\log_2 2}) = O(k)$$

$F(k) \in O(k)$ but we must find $T(n)$

$$\text{if } n = 2^k \quad k = \log_2 n \quad \text{so; } F(k) \in O(\log_2 n)$$

$$\text{and because } T(n) = F(k), \quad \underline{T(n) \in O(\log_2 n)}$$

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

$$T(2) = 2 T(0) = 2^1$$

$$T(3) = 2 T(1) = 2^1$$

$$T(4) = 2 T(2) = 2^2$$

$$T(5) = 2 T(3) = 2^2$$

⋮

$$T(n) = 2 T(n-2) = 2^{\lfloor \frac{n}{2} \rfloor} \quad \text{so; } T(n) \in O(\sqrt{2}^n)$$

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

$$a=4, b=2, d=1$$

$4 > 2^1$ so, case 3

$$T(n) \in \underline{O(n^{\log_2 4})} = \underline{O(n^2)}$$

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

$$\text{lets say } n = 3^k$$

$$T(3^k) = 2 T(3^{k/3}) + 1 \quad \text{and} \quad T(3^k) = F(k)$$

$$F(k) = 2 T(k/3) + 1$$

$$a=2, b=3, d=0$$

$$2 > 3^0 \text{ so case 3; } F(k) \in O(k^{\log_3 2})$$

Now convert $F(k)$ to $T(n)$

$$\text{if } n = 3^k, \quad k = \log_3 n \quad \text{so} \quad F(k) \in O((\log_3 n)^{\log_3 2})$$

$$\log_3 2 = 0.63 \quad \text{and} \quad F(k) = T(3^k) = T(n) \quad \text{so,}$$

$$\underline{T(n) \in O((\log_3 n)^{0.63})}$$

2-) function f(n)

if $n \leq 1$:

print-line ("++")

else

for $i=1$ to n ——— n
 $f(n/2)$ ——— $T(\frac{n}{2})$

end for

So this algorithm $T(n) = n \cdot T(\frac{n}{2})$, $T(1) = 1$

lets solve this by using back

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

$$= n \cdot \frac{n}{2} \cdot T(\frac{n}{4})$$

$$= \frac{n}{2^0} \cdot \frac{n}{2^1} \cdot \frac{n}{2^2} \cdot T(\frac{n}{2^3})$$

$$= \frac{n^k}{2^{\frac{k(k-1)}{2}}} \cdot T(\frac{n}{2^k}) \quad \frac{n}{2^k} = 1, n = 2^k, k = \log_2 n$$

$$T(n) = \frac{n^k}{2^{\frac{k(k-1)}{2}}} \cdot T(1) = \frac{n^{\log_2 n}}{\frac{(\log_2 n - 1) \log_2 n}{2}}$$

Now we need prove this, firstly check for $n=1$

$$T(1) = \frac{1^{\log_2 1}}{\frac{(\log_2 1 - 1) \log_2 1}{2}} = \frac{1^0}{2^0} = 1 \quad \checkmark \text{ True}$$

Assume $T(n)$ is true

$$T(n) = \frac{n^{\log_2 n}}{\frac{(\log_2 n - 1) \log_2 n}{2}}$$

Now prove $T(2n)$ is also true

$$T(2n) = 2n \cdot T(n) = 2n \cdot \frac{n^{\log_2 n}}{\frac{(\log_2 n - 1) \log_2 n}{2}} \quad (1)$$

$$\text{Also } T(2n) = \frac{(2n)^{\log_2 2n}}{\frac{(\log_2 2n - 1) \log_2 2n}{2}}$$

$$= \frac{2n \cdot n^{\log_2 2n}}{\frac{\log_2 n \cdot (\log_2 n + 1)}{2}} = \frac{2n^2 \cdot n^{\log_2 n}}{\frac{\log_2 n (\log_2 n + 1)}{2}} \cdot \frac{n^{-1}}{n^{-1}}$$

$$\text{and } n^{-1} = 2^{\log_2 n^{-1}}$$

$$= \frac{2n \cdot n^{\log_2 n}}{\frac{\log_2 n (\log_2 n + 1)}{2} \cdot 2} = \frac{2n \cdot n^{\log_2 n}}{\frac{\log_2 n (\log_2 n + 1)}{2}} \quad (2)$$

Because of (1) and (2) equal we prove this recurrence. so-

$$T(n) = \frac{n^{\log_2 n}}{\frac{(\log_2 n - 1) \log_2 n}{2}}$$

3) Algorithm function - $F(A[0 \dots n-1])$

if $n=2$ and $A[0] > A[1]$, then swap $(A[0], A[1])$

if $n > 2$ then {

function - $F(A[0 \dots \text{ceil}(\frac{2n}{3})])$

Function - $F(A[\text{floor}(\frac{n}{3}) \dots n])$

Function - $F(A[0 \dots \text{ceil}(\frac{2n}{3})])$

}

This algorithm divide array $\frac{2}{3}$ for every recursion call. So its recursive call 3 times for $\frac{2}{3}$ part of array and sort it so, we can show its recurrence relation as follow;

$$T(n) = T(\lceil \frac{2n}{3} \rceil) + T(\lfloor \frac{2n}{3} \rfloor) + T(\lceil \frac{2n}{3} \rceil) + 1$$

↓
first recursive call
call function for left
 $\frac{2}{3}$ part of array

↓
second recursive
call
call function for
right $\frac{2}{3}$ part
of array

↓
third recursive call
call function again
for left $\frac{2}{3}$ part of
array

So; we can say

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

↓
3 recursive call for
 $\frac{2}{3}$ part of array

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

and, by using master theorem we can solve this recurrence;

$$a=3 \quad b=\frac{3}{2}, \quad d=0$$

$$3 > \frac{3}{2}^0, \quad a > b^d \quad \text{case 3; } O(n^{\log_b a}) = O(n^{\log_{\frac{3}{2}} 3})$$

$$\log_{\frac{3}{2}} 3 \approx 2.71$$

So; this algorithm $T(n) \in O(n^{2.71})$

4-) procedure Insertion Sort($L[1:n]$)

```
for  $i=2$  to  $n$  do
    current  $\leftarrow L[i]$ 
    pos  $\leftarrow i-1$ 
    while (pos  $\geq 1$ ) and (current  $< L[pos]$ )
         $L[pos+1] \leftarrow L[pos]$   $\rightarrow$  swap operation
        pos  $\leftarrow pos-1$ 
    end while
     $L[pos+1] \leftarrow current$ 
end for
end
```

Insertion sort average

We saw average case of insertion sort in class;

Let $T_i = \#$ of basic operation at step i , $1 \leq i \leq n-1$

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

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

Here T_i is a random variable, its average expected number of comparison.

for calculate $E[T_i] = \sum_{j=1}^i j \cdot \underbrace{P(T_i=j)}_{\text{probability that there are } j \text{ comparisons in the } i^{\text{th}} \text{ step}}$

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

$$E[T_i] = \sum_{j=1}^{i-1} j \cdot \frac{1}{i+1} + i \cdot \frac{2}{i+1}$$

$$= \frac{i(i-1)}{2(i+1)} + \frac{2i}{i+1} = \frac{i^2 - i + 4i}{2(i+1)} = \frac{i(i+3)}{2(i+1)}$$

$$= \frac{1}{2} + 1 - \frac{1}{i+1}$$

$$A(n) = E[T] = \sum_{i=1}^{n-1} E[T_i] = \sum_{i=1}^{n-1} \frac{i}{2} + 1 - \frac{1}{i+1}$$

$$= \underbrace{\frac{n(n-1)}{4}}_{\text{This part}} + n - 1 - \underbrace{\sum_{i=1}^{n-1} \frac{1}{i+1}}_{\text{This is a harmonic series so its complexity } O(\log n)}$$

This part
complexity
 $O(n^2)$

This is a harmonic series so its
complexity $O(\log n)$

So, theoretically insertion sort average case complexity $O(n^2)$

procedure quickSort(L[low:high])

if high > low then

call rearrange(L[low:high], position)

call quickSort(L[low:position-1])

call quickSort(L[position+1:high])

endif

procedure rearrange(L[low:high], position)

right = low

left = high + 1

x = L[low]

while (right < left)

repeat right = right + 1 until L[right] ≥ x

repeat left = left - 1 until L[left] ≤ x

if (right < left)

call swap(L[left], L[right])

endif

end while

Also, we learn quick sort average case in class

Lets say;

$$T = T_1 + T_2$$

↓
of
operation
in partition

↘ # of
operation
in
recursive calls

$$A(n) = E[T] = E[T_1] + E[T_2]$$

\swarrow
 this is always
 high-low + 2
 = n+1 operation

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

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

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

$$A(n) = (n+1) + \frac{2}{n} [A(0) + A(1) + \dots + A(n-1)]$$

$$n \cdot A(n) = n \cdot (n+1) + 2 [A(0) + A(1) + \dots + A(n-1)]$$

$$(n-1) \cdot A(n-1) = (n-1) \cdot n + 2 [A(0) + A(1) + \dots + A(n-2)]$$

$$\frac{1}{n(n+1)} [n \cdot A(n) - (n-1) A(n-1)] = \frac{2n + 2A(n-1)}{n(n+1)}$$

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

$$\text{if we say } t(n) = \frac{A(n)}{n+1}$$

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

$$t(0) = 0$$

by using backward substitution

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

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

$$= t(n-3) + \frac{2}{n-1} + \frac{2}{n} + \frac{2}{n+1}$$

$$t(n) = \sum_{i=2}^n \frac{2}{i+1} = 2 \cdot H(n+1) - 3$$

$$A(n) = (n+1) t(n) = 2(n+1) \cdot H(n+1) - 3(n+1) \in O(n \log n)$$

So; Theoretically insertion sort average case $O(n^2)$ and

Quick sort average case $O(n \log n)$. Theoretically Quick sort

average case seems better than insertion sort average case.

Also in my implementation I create 50 random array that has length 10 and sort them with quicksort and Insertion sort.

I find, for quick sort average 18.8 swap operation and 23.3

for insertion sort. So, as we seen in theoretical analyz quicksort

really has a better average case than insertion sort. Nevertheless

insertion sort close to quick sort so it's not a bad algorithm

5-) a) An algorithm that divides the problem into 5 subproblems where the size of each subproblem is one third of the original problem size, solves each subproblem recursively and then combines the solutions to the subproblems in quadratic time.

This algorithm recurrence relation;

$$T(n) = 5 T\left(\frac{n}{3}\right) + n^2$$

↓ ↓ ↘
subproblem one third quadratic time

and we can solve this recurrence relation by using master theorem

$$a=5, \quad b=3, \quad d=2$$

$5 < 3^2$, so because of $a < b^d$ from master theorem

$$T(n) \in \mathcal{O}(n^d) \text{ it means } \underline{T(n) \in \mathcal{O}(n^2) \in \mathcal{O}(n^2)}$$

b) An algorithm that divides the problem into 2 subproblems where the size of each subproblem is half of the original problem size, solves each subproblem recursively and then combines the solutions to the subproblems in $\mathcal{O}(n^2)$ time.

This algorithm recurrence relation;

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

by using master theorem we can solve this recurrence easily

$$a=2, \quad b=2, \quad d=2, \quad 2 < 2^2 \text{ so } a < b^d \text{ case from theorem}$$

$$\text{so } T(n) \in \mathcal{O}(n^d) \in \mathcal{O}(n^2) \text{ so } \underline{T(n) \in \mathcal{O}(n^2)}$$

c) An algorithm that solves the problem by recursively solving the subproblem of size $n-1$ and then combine the solution in linear time

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

Let's use backward method to solve this

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

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

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

$$= T(n-4) + n + n + n + n - 1 - 2 - 3$$

$$\vdots$$
$$= T(n-k) + kn - \left(\frac{k \cdot (k+1)}{2}\right)$$

If $k=n$ then

$$T(n-k) + kn - (k-1) = T(0) + n^2 - \left(\frac{n^2 + n}{2}\right)$$

$$= T(0) + \frac{2n^2 - n^2 - n}{2}$$

$$= T(0) + \frac{n^2 + n}{2}$$

↓
It's a constant

So, $T(n) = \frac{n^2 + n}{2}$, now let's prove this by induction

$$T(0) = 0 \quad \checkmark$$

Assume $T(n-1)$ is true

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

Now prove $T(n)$ is also true

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

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

$$T(n) = \frac{n^2 + n}{2} \text{ so, it is proven } T(n) \text{ is true}$$

Hence, $T(n) \in \Theta(n^2)$ also mean $T(n) \in O(n^2)$

So, both a, b, c has $O(n^2)$ time complexity
we can choose any of them.