Theoretical Assignment 2

Amit Yadav 160099

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Problem 1

Let we are given a pointer to head of a linked list which is to be checked for a cycle. We initialize two pointers with pointing to the head of the linked list.

Pseudo Code:

```
struct node{
           int val;
           struct node* next;
       }NODE;
       void check_cycle(NODE* head){
           NODE* p_one=head->next;;
           NODE* p_two=head->next->next;
            while (1) {
10
               if (p->one=NULL or p->two=NULL or p_two->next=NULL ) {
                   return 0; //no cycle
12
13
               if(p_one = p_two)
14
                   return 1; //cycle present
16
               p\_one=p\_one->next; //shift by one node
17
               p_two=p_two->next->next; //shift by two nodes
18
19
20
21
            }
22
```

Termination

- If linked list doesnot contain any cycle, then code will terminate when p_one or p_two reaches NULL pointer.
- If cycle is present, then the program will terminate when p_one and p_two point to same node. Since, the distance between p_one and p_two decreases by one in every iteration and maximum initial distance can be equal to the size of the cycle, so it's certain that they will meet/point to same node before p_one reaches the end of the cycle.

Time complexities

1. **Big Oh(O):** Since the loop will always terminate before the p_one pointer reaches end of the cycle, so the loop will iterate at most N times (size of the linked list). Hence

$$time = O(n)$$

2. Big omega(Ω): If the cycle includes all elements/nodes of the linked list, the loop will iterate at N or N+1 times. Else, it will take less then N iteration, therefore

time =
$$\Omega(n)$$

3. Big Theta(Θ): Since time complexity is in both O(n) and $\Omega(n)$, implies time is also in $\Theta(n)$

$$time = \Theta(n)$$

Problem 2

1. First loop iterates N times. And 2nd nested loop iterates 'i' times for i in $\mathrm{range}(0,N)$ with N excluded.

i	0	1	2	3		N-1
j		0	0	0		0
			1	1		1
				2		2
						N-2

So total number of executions of statement

$$sum+=j;$$

is

$$\sum_{i=0}^{n-1} i = O(n^2)$$

2. Our inner loop will run till nearest smaller power of 2 times i.e $2^{floor(\log_2(N-1))}$ times. For ex. if N=16, then $2^{floor(\log_2 15)} = 8$. So i will range in (0..8).

i	1	2	4	8			$2^{floor(\log_2(N-1))}$
j	0	0	0	0			0
		1	1	1			1
			2	2			2
			3				•
				7			
						٠	$2^{floor(\log_2(N-1))} - 1$

Therefore,

$$Time = \sum_{i=1}^{\log_2(N-1)} 2^i$$

For ex. if N=16, time = 1+2+4+8=15 units = N-1 Hence,

$$time = O(n)$$

3. Outer loop iterates $\log_2(N-1)$ times. While for every iteration, nested loop runs $\log_2(i)$ times.

i	1	2	4	8		N
j		1	1	1		1
			2	2		2
				4		4
						2^{-1}

$$Time = \sum_{i=1}^{\log(N)-1} i$$

i.e

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

4. Let our computer can store smallest positive real number = y. Then after x iterations,

$$i = \frac{N}{2^x} > y$$

$$\log N = \log y + x$$

$$x = O(\log N)$$

Hence, $time = O(\log n)$

1. $a \times \log n$ and $\log_a n$

$$f(n) = a \log n$$
$$g(n) = \log_a n = \frac{\log n}{\log a}$$

Let

$$c \times \frac{\log n}{\log a} \ge a \times \log n$$

It holds for $c = 2 \times a \times \log a$ and $n \ge 1$ Hence,

$$f(n) = O(g(n))$$

2. 7^{4n} and $2^{n/11}$

$$c \times 2^{n/11} \ge 7^{4n}$$

$$c \ge \frac{7^{4n}}{2^{n/11}} \ge \frac{7^{4n}}{2^n} \ge \frac{7^{4n}}{7^n} = 7^{3n}$$

This means, when $n \to \infty$, c should be infinity, which is not possible. Therefore

$$f(n) \neq O(g(n))$$

3. $2^{\sqrt{\log n}}$ and \sqrt{n}

$$2^{\sqrt{\log n}} = 2^{\frac{\log n}{\sqrt{\log n}}} = (2^{\log n})^{\frac{1}{\sqrt{\log n}}} = n^{\frac{1}{\sqrt{\log n}}}$$

Now,

$$f(n) = n^{\frac{1}{\sqrt{\log n}}}$$
 and $g(n) = \sqrt{n}$

Clearly,

$$c \times g(n) \ge f(n) \ \forall n \ge 2^4 \ and \ c = 1$$

Therefore,

$$f(n) = O(g(n))$$

4. $\sum_{i=1}^n \frac{1}{i}$ and $\log n$ Both functions are strictly increasing with decreasing slope. Let's see which of them is increasing faster. We calculate f(n+1)-f(n) and g(n+1)-f(n)1) - g(n)

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

because $\frac{d(\log n)}{dn} = \frac{1}{n}$ and it's slope is decreasing. Now, clearly g(n) is increasing faster than f(n). Therefore,

$$f(n) = O(g(n))$$

1.
$$T(n)=c+T(n/k)$$

$$T(n)=c+T(n/k)$$

$$T(n)=c+c+T(n/k^2)$$

$$T(n)=c+c+\ldots+c_{i-times}+T(n/k^i)$$

$$T(n)=c\times\log_k n+T(1)$$

Therefore, time complexity of T(n) is

$$time = O(\log_k n) = O(\log n)$$

$$2. \ T(n) = c + T(n^{1/\alpha})$$

$$T(n) = c + T(n^{1/\alpha})$$

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

$$T(n) = c + c + c + \dots + c + T(n^{1/\alpha^i})$$

$$T(n) = c \times \frac{\log(\log_2 n)}{\log \alpha} + T(x)$$

Because $n^{1/\alpha^i} = 2$

$$T(n) = c \times \frac{\log(\log_2 n)}{\log \alpha} + c$$

Therefore. time complexicity of T(n) is

$$time = O(\log \log n)$$

3.
$$T(n) = c \times n + T(n/k)$$

$$T(n) = c \times n + \frac{c \times n}{k} + \frac{c \times n}{k^2} + \dots + \frac{c \times n}{k^i} + T(n/k^i)$$

$$T(n) = c \times n + \frac{c \times n}{k} + \frac{c \times n}{k^2} + \dots + \frac{c \times n}{k^i} \log_k n - times$$

$$T(n) = cn \times (1 + 1/k + 1/k^2 + \dots) \log_k n - times$$

$$T(n) = cn \times \frac{(1 - k^{\log_k 1/n})}{1 - 1/k}$$

$$T(n) = O(n)$$

1. 4n+7 is o(n): Flase

For 4n+7 to be in o(n), $4n+7 < K \times n$ for all k > 0 and some $n > n_0$, which is not the case here. If k=1, we can't find any n_0 which satisfy this condition.

2. 4n+7 is $o(n^2)$: True $4n+7 < k \times n^2$ for all k>0 and $n>n_0$ since

$$\lim_{n \to \infty} \frac{4n+7}{n^2} = 0$$

3. 4n + 7 is $\omega(n)$: Flase

Similiar to first part. We can't find n_0 such that $4n + 7 > k \times n \forall n > n_0$ and k = 5 or greater.

4. 4n + 7 is $\omega(\log n)$: True

$$\lim_{n \to \infty} \frac{4n+7}{\log n} \ is \ \infty$$

Therefore, $\log n$ is an loose lower bound of 4n+7.

Problem 6

- 1. Depends on the values of c and k
 - (a) f(n) is O(g(n)): Possible if $c \le 1$ and k > 1
 - (b) f(n) is $\Omega(g(n))$: True for all values of c > 1
 - (c) f(n) is $\Theta(g(n))$: Only possible if c=1 and k=0
- 2. $f(n) = \log_2 n, g(n) = \ln(n)$:
 - (a) f(n) is O(g(n)): True. $c \times \log_e n \ge \log_2 n$ for $c \ge \log_2 e$ and $n \ge 1$
 - (b) f(n) is $\Omega(g(n))$: True. $c \times \log_e n \le \log_2 n$ for $c \le \log_2 e$ and $n \ge 1$
 - (c) f(n) is $\Theta(g(n))$: Since f(n) is in both O(g(n)) and $\Omega(g(n))$, so f(n) is in $\Theta(g(n))$.

This can also be done by $\lim_{n\to\infty} \frac{f(n)}{g(n)}$ which is finite.

3. $f(n) = n^2 \log_2 n$, $g(n) = n \log_2 n^3$

$$g(n) = n \log_2 n^3 = 3n \log_2 n$$

$$\lim_{n \to \infty} \frac{f(n)}{g(n)} \text{ is } \infty$$

Therefore,

$$f(n) = \Omega(g(n))$$

Instead of trivial iterative solution, we can use divide and conquer method. We follow the same algorithm of merge sorting and also keep counting the number of inversions.

Pseudo code:

```
int count=0; //number of inversions
        void sort(int *a,int left, int right){
              if (left < right) {</pre>
                   mid = (left + right)/2;
                   sort(int *a, left, mid)
                   sort(int *a, mid+1, right)
                   merge(int *a, left, mid, right)
                   return;
              }
9
              else return;
10
        merge(int *a, left, mid, right){
13
               \begin{array}{lll} \textbf{int} & \textbf{i} \!=\! \textbf{left} \ , & \textbf{j} \!=\! \textbf{mid} \!+\! 1, & \textbf{k} \!=\! 0; \end{array} 
14
              int b[right-left+1];
15
16
              while (i \leq\ mid && j \leq\ right) {
                   if (a[i] < a[j]) {
                        b[k++]=a[i++];
                   }
19
20
                        count=count+(mid-i+1); //if element of left array
21
        // is larger, then there will be (size_of_array-i) inversions
                        b[k++]=a[j++];
23
24
              while (i \le mid) b [k++]=a[i++];
25
              while (j \le right) b [k++]=a[j++];
26
27
              for (k=0, i=left; k < right - left + 1; k++)
28
29
                   a[i++]=b[k];
30
31
              return;
```

Time comlexity:

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

$$T(n) = 2[2T(\frac{n}{4}) + \frac{n}{2}] + n$$

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

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

$$T(n) = nT(1) + n\log n$$

$$\mathbf{T(n) = O(n\log n)}$$