

COMP 3711

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1 Tutorials

1.1 Tutorial 1 - Asymptotic notation

1. (a) true
(b) false
(c) true
(d) false
(e) false
(f) false
(g) true
2. (a) true
(b) false
(c) false
3. (a) yes
(b) no.

let $T_i(n) = i \cdot n$ and $f(n) = n$

$$\begin{aligned} g(n) &= \frac{n(n+1)}{2} \cdot f(n) \\ &= O(n^2 f(n)) \\ &= O(n^3). \end{aligned}$$

$\therefore g(n) \neq O(f(n))$ and $\neq O(n(f(n)))$.

note: this is because

$$\sum_{i=1}^n O(f(n)) \neq O\left(\sum_{i=1}^n f(n)\right).$$

1.2 Tutorial 2 - Divide and conquer

1. done
2. done
3. find the median. If $k > \text{median}$, then remove all i from A s.t where $A[i] < \text{median}$, vice versa. Time complexity: $\frac{n}{2} + \frac{n}{4} + \dots + \frac{n}{n} = \frac{(\frac{1}{2})^{\log_2 \frac{n}{2}-1}}{\frac{1}{2}-1} \cdot \frac{n}{2} = \frac{\frac{1}{2}-1}{\frac{1}{2}-1} \cdot \frac{n}{2} = \frac{1-\frac{n}{2}}{\frac{1}{2}-1} = O(n)$
4. find the median x_k and then $w = \sum_{i=k}^n w_k$. W.L.O.G, if $w > \frac{1}{2}$, then our solution $j > k$. Remove $A[i]$ such that $i < k$. Set $w_k = w_k + 1 - w$. If $w < \frac{1}{2}$, remove all $A[j]$ where $j > k$. Set $w_k = w$. Continue the recursion. The time complexity will be $O(n)$ (analysis the same as Q3)

1.3 Tutorial 3 - Divide and conquer

1. done
2. take a pivot. check which side is larger. Take the larger side. If all elements in one side is the same and $\text{len} > \frac{n}{2}$ then we have our answer. Recurse until both sides have $\text{len} < n/2$
3. example: 5, -10, -10, 0, 60

2 Lectures

2.1 Lecture 1 - Asymptotic Notation

Theorem 2.1

Upper bounds $T(n) = O(f(n))$.

if exists constants $c > 0$ and n_0 such that $\forall n \geq n_0, T(n) \leq c \cdot f(n)$

Theorem 2.2

Lower bounds $T(n) = \Omega(f(n))$

if exists constants $c > 0$ and n_0 such that $\forall n \geq n_0, T(n) \geq c \cdot f(n)$

Theorem 2.3

Tight bounds $T(n) = \Theta(f(n))$

if $T(n) = O(f(n))$ and $T(n) = \Omega(f(n))$

constant $9999^{9999^{9999}} < \text{logarithmic } \log^{10} n < \text{polynomial } n^{0.1} < n \log n < n^2 < \text{exponential } 2^n$

$$1. \ 9999^{9999^{9999}} = \Theta(1) = O(\log(\log(n)))$$

$$2. \ \log(\log(n)) = O(\log n)$$

$$3. \ n^{100} = O(2^n) \text{ Let } n^{100} = c \cdot 2^n \text{ s.t. for } n \geq n_0 n^{100} \leq c \cdot 2^n$$

$$100 * \log(n) = c \cdot n \log(2)$$

$$n = \frac{100}{c} * \log(n)$$

.

$$\therefore \forall c, n^{100} = O(2^n)$$

Theorem 2.4

Common expressions

1. $\max(f(n), g(n)) = \Theta(f(n) + g(n))$
2. $\log \sqrt{n} = \Omega(\sqrt{\log(n)})$
3. $\log(2^n) = \Theta(\log(3^n))$
4. $\sum_{i=1}^n \frac{1}{i} = \Theta(\log n)$

2.2 Lecture 2 - Running time of sorting

2.2.1 Selection sort

Algorithm 1 Selection sort

```

for  $i \leftarrow 1$  to  $n$  do
  for  $j \leftarrow i + 1$  to  $n$  do
    if  $A[j] < A[i]$  then
      swap  $A[j], A[i]$ 
    end if
  end for
end for
end for

```

number of comparisons : $\sum_{i=1}^n (n - i) = \sum_{i=1}^{n-1} i = \frac{n(n-1)}{2}$

Theorem 2.5

Correctness of selection sort

Prove by induction. Assume the algorithm sorts every array of size $n-1$ correctly.

1. It first puts the smallest item in $A[1]$
2. then runs selection sort on $A[2 \dots n]$ (by induction this is correct)
3. since $A[1]$ is smaller than every other items, the array is sorted

2.2.2 Insertion sort

Algorithm 2 Insertion sort

```

for  $i \leftarrow 2$  to  $n$  do
   $j \leftarrow i - 1$ 
  while  $j \geq 1$  and  $A[j] > A[j + 1]$  do
    swap  $A[j]$  and  $A[j + 1]$ 
     $j \leftarrow j - 1$ 
  end while
end for

```

1. number of comparisons : $\sum_{i=1}^n (i - 1) = \sum_{k=1}^{n-1} k = \frac{n(n-1)}{2}$
2. average case analysis: only half of the keys are compared
3. best case: on sorted data, takes $O(n)$ time

Theorem 2.6

Comparison of running time

1. $O(\log n) \cup \Theta(2^{\log_2 \log_2 n})$
2. $O(n^4) \cup O(n^3)$

Exercise 2.7Prove that $\log(n!) = \Theta(n \log n)$

1. first, prove $\log(n!) = O(n \log n)$

$$\begin{aligned}\log(n!) &= \sum_{i=1}^n \log(i) \\ &\leq \sum_{i=1}^n \log(n) \\ &= O(n \log n).\end{aligned}$$

2. then, prove $\log(n!) = \Omega(n \log n)$

$$\begin{aligned}\log(n!) &= \sum_{i=1}^n \log(i) \\ &\geq \sum_{i=\frac{n}{2}}^n \log(i) \\ &\geq \sum_{i=\frac{n}{2}}^n \log\left(\frac{n}{2}\right) \\ &= \frac{n}{2} \log \frac{n}{2} \\ &= \frac{n}{2} (\log n - \log 2) \\ &= \Omega(n \log n).\end{aligned}$$

$$\therefore \log(n!) = \Theta(n \log n)$$

2.3 Lecture 3 - Divide and conquer

Theorem 2.8

Run time analysis of Binary search

$$\begin{aligned}
 T(n) &= T\left(\frac{n}{2}\right) + 2 \text{ if } n > 1, \text{ with } T(1) = 1 \\
 &= T\left(\frac{n}{2^2}\right) + 2 \\
 &= \dots \\
 &= T\left(\frac{n}{2^{\log_2 n}}\right) + 2 \log_2 n \\
 &= 1 + 2 \log_2 n.
 \end{aligned}$$

Examples

1. Rotated sorted array
2. Find the last 0

Algorithm 3 Tower of hanoi(n, peg1, peg2, peg3)

```

1: if n=0 then return
2: else
3:   Tower of hanoi(n-1, peg1, peg2, peg3)           ▷ T(n-1)
4:   move the only disc from peg 1 to peg 3           ▷ T(1)
5:   Tower of hanoi(n-1, peg2, peg1, peg3)           ▷ T(n-1)
6: end if

```

Theorem 2.9

Recurrence of Tower of hanoi:

$$\begin{aligned}
 T(n) &= 2T(n-1) + 1 \\
 &= 2(2T(n-2) + 1) + 1 \\
 &= \dots \\
 &= 2^{n-1}T(1) + 2^{n-2} + 2^{n-3} + \dots + 2^2 + 2 + 1 \\
 &= 2^n - 1.
 \end{aligned}$$

Algorithm 4 Merge sort(A, l, r)

```

1: if  $l = r$  then return
2: else
3:    $\text{mid} \leftarrow \frac{l+r}{2}$ 
4:   Merge Sort(A, l, mid)  $\triangleright T(\frac{n}{2})$ 
5:   Merge Sort(A, mid+1, r)  $\triangleright T(\frac{n}{2})$ 
6:   Merge(A, l, mid, r) Comment  $O(n)$ 
7: end if

```

Theorem 2.10

Analysis of merge sort

$$\begin{aligned}
 T(n) &\leq T\left(\left\lfloor \frac{n}{2} \right\rfloor\right) + T\left(\left\lceil \frac{n}{2} \right\rceil\right) + O(n) \\
 &= 2^i T\left(\frac{n}{2^i}\right) + in \quad \text{where } i = \log_2 n \\
 &= n \log_2 n + n.
 \end{aligned}$$

2.3.1 Inversion number**Definition 2.11.** Inversion number

Given an array $A[1..n]$, two elements $A[i]$ and $A[j]$ are inverted if $i < j$ but $A[i] > A[j]$

Algorithm 5 Count number of inversion**Input:** A, p, q, r **Output:** c

```

1:  $L \leftarrow A[p \dots q]$  and  $R \leftarrow A[q \dots r]$ 
2:  $c \leftarrow 0$ 
3: while  $i \leftarrow 0 \leq p - q + 1$  and  $j \leq r - q$  do
4:   if  $L[i] \leq R[j]$  then
5:      $i \leftarrow i + 1$ 
6:   else
7:      $I[j] = q - p - i + 2$ 
8:      $c \leftarrow c + I[j]$ 
9:      $j \leftarrow j + 1$ 
10:  end if
11: end while
12: return  $c$ 

```

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

Algorithm 6 Sort and count

Input: A, p, r **Output:** c

```
1: if  $p = r$  then
2:   return 0
3: end if
4:  $c_1 \leftarrow \text{SORT-AND-COUNT}(A, p, q)$ 
5:  $c_2 \leftarrow \text{SORT-AND-COUNT}(A, q + 1, r)$ 
6:  $c_3 \leftarrow \text{MERGE-AND-COUNT}(A, p, q, r)$ 
7: return  $c_1 + c_2 + c_3$ 
```

Algorithm 7 Merge and count

Input: A, p, q, r **Output:** c

```
1:  $L \leftarrow A[p \dots q]$  and  $R \leftarrow A[q \dots r]$ 
2:  $c \leftarrow 0$ 
3: for  $k \leftarrow p$  to  $r$  do
4:   if  $L[i] \leq R[i]$  then
5:      $A[k] \leftarrow L[i]$ 
6:      $i \leftarrow i + 1$ 
7:   else
8:      $A[k] \leftarrow R[j]$ 
9:      $I[j] = q - p - i + 2$ 
10:     $c \leftarrow c + I[j]$ 
11:     $j \leftarrow j + 1$ 
12:   end if
13: end for
14: return  $c$ 
```

2.3.2 Subarray

1. Maximum subarray (DC v.s. Kadane's algorithm)

Algorithm 8 Kadane's algorithm**Input:** *input***Output:** *output*

```

1:  $V \leftarrow 0$ 
2:  $maxi \leftarrow -\infty$ 
3: for  $i \leftarrow 1$  to  $n$  do
4:    $V \leftarrow V + A[i]$ 
5:   if  $V < A[i]$  then
6:      $V = A[i]$ 
7:   end if
8:   if  $V > maxi$  then
9:      $maxi = A[i]$ 
10:  end if
11: end for
12: return state

```

2.3.3 Integer multiplication**Theorem 2.12**

Karatsuba Multiplication

Let $a = a_1 2^{\frac{n}{2}} + a_0$ and $b = b_1 2^{\frac{n}{2}} + b_0$

$$ab = a_1 b_1 2^n + (a_1 b_0 + a_0 b_1) 2^{\frac{n}{2}} + a_0 b_0$$

$$a_1 b_0 + a_0 b_1 = (a_0 + a_1) \cdot (b_0 + b_1) - a_0 b_0 - a_1 b_1 \quad (\text{reduce to 3 multiplications})$$

.

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

2.4 Lecture 4 - Master theorem**2.4.1 intro**

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

1. $f(n) = n^{c+\epsilon}$ then $T(n) = \Theta(n^c)$
2. $f(n) = n^c$ then $T(n) = O(n^c)$
3. $f(n) = n^{c-\epsilon}$, then $T(n) = \Omega(f(n))$

2.4.2 inequality

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

1. $f(n) = O(n^{c+\epsilon})$ then $T(n) = O(n^c)$

2. $f(n) = O(n^c)$ then $T(n) = O(n^c)$
3. $f(n) = O(n^{c-\epsilon})$, then $T(n) = O(f(n))$

Remark 2.13 — when $f(n) = \Theta(n)$, we will split the cases by checking $c > 1$, $c = 1$ and $c < 1$

Proof when $f(n) = O(n)$

1.

$$\begin{aligned}
 T(n) &\leq aT\left(\frac{n}{b}\right) + f(n) \\
 &= a \cdot \left(T\left(\frac{n}{b^2}\right) + \left(\frac{n}{b}\right)\right) + f(n) \\
 &= \dots \\
 &= a^i T(1) + a^{i-1} \left(\frac{n}{b^i}\right) + \dots + kn \\
 &= n^c + kn \sum_{i=0}^{\log_b(n)-1} \left(\frac{a}{b}\right)^i && \text{case 1: } c > 1 \\
 &= n^c + kn \cdot \frac{\left(\frac{a}{b}\right)^{\log_b n} - 1}{\frac{a}{b} - 1} \\
 &\leq n^c + kn \cdot \frac{\left(\frac{a}{b}\right)^{\log_b n}}{\frac{a}{b} - 1} \\
 &= n^c + k \cdot \frac{n^{\log_b a}}{\frac{a}{b} - 1} \\
 &= O(n^c) \\
 &\quad .
 \end{aligned}$$

2.

$$\begin{aligned}
 T(n) &\leq aT\left(\frac{n}{b}\right) + f(n) \\
 &= n + kn \sum_{i=0}^{\log_b(n)-1} \left(\frac{a}{b}\right)^i && \text{case 2: } c = 1 \\
 &= n + kn \cdot (\log_b n - 1) \\
 &= O(n \log n) \\
 &\quad .
 \end{aligned}$$

3.

$$\begin{aligned}
T(n) &\leq aT\left(\frac{n}{b}\right) + f(n) \\
&= n^c + kn \sum_{i=0}^{\log_b(n)-1} \left(\frac{a}{b}\right)^i \quad \text{case 3: } c < 1 \\
&= n^c + kn \cdot O(1) \quad \text{decreasing geometric series} \\
&= O(n)
\end{aligned}$$

2.5 Lecture 5 - Randomized algorithms

Exercise 2.14

$$\begin{aligned}
S &= \sum_{i=1}^{\infty} i(1-p)^{i-1} = \sum_{j=0}^{\infty} (j+1)(k)^j \\
kS &= \sum_{j=1}^{\infty} (j+1)(k)^j \quad (1)
\end{aligned}$$

$$S - kS = 1 + k + k^2 + k^3 + \dots = \frac{1}{1-k} \quad (2)$$

$$S = \frac{1}{(1-k)^2}$$

Exercise 2.15 (Coupon collector)

$$\begin{aligned}
P(i) &= \frac{n-i}{n} \\
E[x_i] &= \frac{1}{P(i)} \\
E[X] &= \sum_{i=1}^{n-1} E[x_i] \\
&= n \sum_{i=1}^n \frac{1}{i} \\
&= \Theta(n \log n)
\end{aligned}$$

1. hiring problem
2. birthday paradox
3. generate random permutation (with proof)

2.6 Lecture 6 - Quick Sort

2.6.1 Intro to algo

Algorithm 9 Quicksort

Input: A, p, q, r **Output:** 1

```
1: if  $p \geq r$  then
2:   return
3: end if
4:  $q = \text{PARTITION}(A, p, r)$ 
5:  $\text{QUICKSORT}(A, p, q - 1)$ 
6:  $\text{QUICKSORT}(A, q + 1, r)$ 
7: return state
```

Algorithm 10 Partition

Input: A, p, r **Output:** q

```
1:  $i \leftarrow p - 1$ 
2: for  $j \leftarrow p$  to  $r - 1$  do
3:   if  $A[j] \leq x$  then
4:      $i \leftarrow i + 1$ 
5:      $\text{SWAP}(A[i], A[j])$ 
6:   end if
7: end for
8: return  $i$ 
```

2.6.2 Run time analysis

1. best case: $O(\log n)$
2. worse case: $O(n^2)$

Average run-time of quicksort. To attain average case, randomly select

the pivot. Let $p(x_{i,j})$ be the probability that z_i and z_j are compared

$$p(x_{i,j}) = \frac{1}{j - i + 1}$$
$$E[X] = \sum_{i=1}^n 2 \cdot p(x_{i,j}) \quad z_i \text{ and } z_j \text{ are symmetric}$$
$$= O(n \log n)$$

Exercise 2.16 (k-smallest element in union of two sorted lists)