

0.1 Lecture 1 - Asymptotic Notation

Theorem 0.1

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 0.2

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 0.3

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 0.4

Common expressions

$$1. \max(f(n), g(n)) = \Theta(f(n) + g(n))$$

$$2. \log \sqrt{n} = \Omega(\sqrt{\log(n)})$$

$$3. \log(2^n) = \Theta(\log(3^n))$$

$$4. \sum_{i=1}^n \frac{1}{i} = \Theta(\log n)$$

0.2 Lecture 2 - Running time of sorting

0.2.1 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

```

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

Theorem 0.5

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

0.2.2 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 0.6

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 0.7Prove 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 $\log(n!) = \Omega(n \log n)$

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

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

0.3 Lecture 3 - Divide and conquer

Theorem 0.8

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 0.9

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 0.10

Analysis of merge sort

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

0.3.1 Inversion number**Definition 0.11.** 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  $c$ 
```

0.3.2 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

```

0.3.3 Integer multiplication**Theorem 0.12**

Karatsuba Multiplication

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

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

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

.

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