Analysis of Algorithm Lecture-04: Merge Sort and Selection Problem

Sorting

- Sorting is a well studied problem from analysis point of view
- Sorting is one of few problems where provable lower bounds exist on how fast we can sort
- In sorting we are given an array A[1..n] of n numbers
- We are to reorder these elements into increasing (or decreasing) order

Divide-and-Conquer

- The Divide-and-Conquer strategy is employed to solve large number of computational problems.
	- Divide: the problem into a small number to pieces.
	- Conquer: solve each piece by applying divide and conquer to it recursively.
	- Combine: the pieces together into a global solution.

Merge Sort

- Divide: Split A down the middle into two subsequences, each of size roughly n/2
- Conquer: Sort each subsequence by calling merge sort recursively on each
- Combine: Merge the two sorted subsequences into a single sorted list.

Merge Sort

- The dividing process ends when we have split the subsequences down to a single item.
- A sequence of length 1 is trivially sorted.


```
MERGE-SORT Algorithm
MERGE-SORT(array A, int p, int r)
if (p < r) then
 q \leftarrow (p+r)/2MERGE-SORT(A, p, q)
 MERGE-SORT(A, q+1, r)MERGE(A, p, q, r)
```
MERGE Algorithm

```
MERGE(array A, int p, int q, int r)
  int B[p..r]; int i \leftarrow k \leftarrow p; int j \leftarrow q+1;
  while (i \le q) and (j \le r)
  do 
        if (A[i] < A[i]) then
                 B[k++] \leftarrow A[i++]:
        else 
                 B[k++] \leftarrow A[i++];
  while (i \leq q) do B[k++] \leftarrow A[i++];
  while (i \le r) do B[k++] \leftarrow A[i++];
  for i \leftarrow p to r do A[i] \leftarrow B[i];
```
Logarithmic identities

$$
y = \log_b(x)
$$
if and only if $x = b^y$
\n
$$
\log_b(xy) = \log_b(x) + \log_b(y)
$$

\n
$$
\log_b\left(\frac{x}{y}\right) = \log_b(x) - \log_b(y)
$$

\n
$$
\log_b(x^y) = y \log_b(x)
$$

\n
$$
\log_b(\sqrt[y]{x}) = \frac{\log_b(x)}{y}
$$

\n
$$
x^{\log(y)} = y^{\log(x)}
$$

\n
$$
\log_b(b) = 1
$$

\n
$$
\log_b(1) = 0
$$

10

Analysis of Merge Sort

- First consider the running time of procedure MERGE(A, p, q, r).
- Let $n = r p + 1$ denote the total length of both left and right sub arrays, i.e. sorted pieces.
- The MERGE procedure contains four loops, none nested in the other.
- Each loop can execute at most n times.
- Let T(n) denote the worst case running time of MERGE-SORT on an array of length n.
- If we call MERGE-SORT with an array containing a single item (n $=$ 1) then the running time is constant.
- We can just write $T(n) = 1$, ignoring all constants.

Analysis of Merge Sort (cont'd)

- For $n > 1$ MERGE-SORT Splits into two halves, sorts the two and then merges them together.
- The left half is of size $\lceil n/2 \rceil$ and the right half is $\lfloor n/2 \rfloor$
- How long does it take to sort the elements in sub array of size
- We do not know this, $\lceil n/2 \rceil$ but because, $\lceil n/2 \rceil < n$ for $n > 1$

we can express this as

$$
T(\left\lceil n/2\right\rceil)
$$

Analysis of Merge Sort (cont'd)

- Similarly the time taken to sort right sub array is expressed as $T(\lfloor n/2 \rfloor)$
- In conclusion we have

$$
T(n) = \begin{cases} 1 & \text{if } n = 1 \\ T(\lceil n/2 \rceil) + T(\lfloor n/2 \rfloor) + n & \text{otherwise} \end{cases}
$$

 This is called recurrence relation i.e. a recursively defined function.

Solving the Recurrence

• Lets expand the terms
\n
$$
T(1) = 1
$$
\n
$$
T(2) = T(1) + T(1) + 2 = 1 + 1 + 2 = 4
$$
\n
$$
T(3) = T(2) + T(1) + 3 = 4 + 1 + 3 = 8
$$
\n
$$
T(4) = T(2) + T(2) + 4 = 8 + 8 + 4 = 12
$$
\n
$$
T(5) = T(3) + T(2) + 5 = 8 + 4 + 5 = 17
$$
\n...\n
$$
T(8) = T(4) + T(4) + 8 = 12 + 12 + 8 = 32
$$
\n...\n
$$
T(16) = T(8) + T(8) + 16 = 32 + 32 + 16 = 80
$$
\n...\n
$$
T(32) = T(16) + T(16) + 32 = 80 + 80 + 32 = 192
$$

14

Solving the Recurrence (cont'd)

- What is the pattern here?
- Let's consider the ratio of $T(n)/n$ for powers of 2

• This suggests $T(n)/n = \log n + 1$ $T(32)/32 = 6$ $T(16)/16 = 5$ $T(8)/8=4$ $T(4)/4=3$ $T(2)/2 = 2$ $T(1)/1 = 1$

Solving the Recurrence (cont'd)

- Floor and ceiling are difficult to deal with
- If n is assumed to be a power of 2 $(2^k = n)$ this will simplify the recurrence to

$$
T(n) = \begin{cases} 1 & \text{if } n = 1 \\ 2T(n/2) + n & \text{otherwise} \end{cases}
$$

The Iteration Method

 $n = 16T(n/16) + n + n + n + n$ $= 8(2T(n/16) + n/8) + n + n + n$ $= 8T(n/8) + n + n + n$ $= 4(2T(n/8) + n/4) + n + n$ $=4T(n/4)+n+n$ $= 2(2T(n/4) + n/2) + n$ $T(n) = 2T(n/2) + n$

The Iteration Method (cont'd)

If n is power of 2 then let $n = 2^k$ or $k = \log n$.

$$
T(n) = 2k T(n/(2k)) + (n+n+n+...+n)
$$

\n
$$
\underbrace{k \text{ times}}
$$

$$
= 2k T(n/(2k)) + kn
$$

= 2^(logn) T(n/(2^(logn))) + (log n)n
= 2^(logn) T(n/n) + (log n)n
= nT(1) + n log n = n + n log n

MERGE-SORT Recursion tree

Example-1: Solving Recurrence

• Assume n to be power of 4 i.e. $n = 4^k$ and $k = log_4 n$

$$
T(n) = \begin{cases} 1 & \text{if } n = 1 \\ 3T(n/4) + n & \text{otherwise} \end{cases}
$$

• Iteration Method

 $27T(n/64) + 9(n/16) + 3(n/4)$ $9T(n/16) + 3(n/4)$ $3(3T(n/16)) + (n/4))$ $(n) = 3T(n/4)$ $T = 27T(n/64) + 9(n/16) + 3(n/4) + n$ $T = 9T(n/16) + 3(n/4) + n$ $= 3(3T(n/16)) + (n/4)) + n$ $T(n) = 3T(n/4) + n$

...

 $\overline{}$

$$
T(n) = 3^{k} T(\frac{n}{4^{k}}) + 3^{k-1}(n/4^{k-1}) + ... + 9(n/16) + 3(n/4) + n
$$

\n
$$
T(n) = 3^{k} T(\frac{n}{4^{k}}) + \sum_{i=0}^{k-1} \frac{3^{i}}{4^{i}}n
$$

\nWith $n = 4^{k}$ and $T(1) = 1$
\n
$$
T(n) = 3^{\log_{4} n} T(1) + \sum_{i=0}^{\log_{4} n} \frac{3^{i}}{4^{i}}n
$$

n remains constant

we thus have throughout the sum and $3' / 4' = (3/4)$; *i i i*

$$
T(n) = n^{\log_4 3} + n \sum_{i=0}^{(\log_4 n) - 1} \left(\frac{3}{4}\right)^i
$$

we used the formula $a^{\log_b n} = n^{\log_b a}$

is ^a geometric series; 4 3 The sum $(\log_4 n) - 1$ 0 4 \int $\bigg)$ $\overline{}$ \setminus $\sum^{g_4 n)-1}$ $=$ *n*)-1 \wedge \rightarrow \vee ^{*i*} *i*

for $x \neq 1$

$$
\sum_{i=0}^{m} x_i = \frac{x^{m+1} - 1}{x - 1}
$$

23

$$
\sum_{i=0}^{m} x_i = \frac{x^{m+1} - 1}{x - 1}
$$

In this case $x = 3/4$ and $m = \log_4 n - 1$ We ge^t

$$
T(n) = n^{\log_4 3} + n \frac{(3/4)^{(\log_4 n) - 1 + 1} - 1}{(3/4) - 1}
$$

 $(3/4)^{\log_4 n} = n^{\log_4(3/4)} = n^{\log_4 3 - \log_4 4}$ Applying the log identity once more

$$
= n^{\log_4 3 - 1} = \frac{n^{\log_4 3}}{n}
$$

if we plug this back, we ge^t

$$
T(n) = n^{\log_4 3} + n \frac{n}{(3/4) - 1}
$$

= $n^{\log_4 3} + \frac{n^{\log_4 3} - n}{-1/4}$
= $n^{\log_4 3} + \frac{n^{\log_4 3} - n}{-1/4} = n^{\log_4 3} + \frac{(n^{\log_4 3} - n)/1}{-1/4} = n^{\log_4 3} + \frac{4(n^{\log_4 3} - n)}{-1} = n^{\log_4 3} + 4(-n^{\log_4 3} + n)$
= $4n - 4n^{\log_4 3} + n^{\log_4 3}$
= $4n - 3n^{\log_4 3}$
with $\log_4 3 \approx 0.79$, we finally have the result!
 $T(n) = 4n - 3n \log_4 3 \approx 4n - 3n^{\frac{0.79}{9}} \in \Theta(n)$

Example-2: Solving Recurrence

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

= $f(n-2) + n - 1 + n$
= $f(n-3) + n - 2 + n - 1 + n$
= $f(n-4) + n - 3 + n - 2 + n - 1 + n$
...
= $f(1) + 2 + 3 + 4 + ... + n - 1 + n$
= $\frac{n(n+1)}{2} = \Theta(n^2)$

26

Example-3: Solving Recurrence

… $=f(n/16) + 1 + 1 + 1 + 1$ $=f(n/8)+1+1+1$ $=f(n/4)+1+1$ $f(n) = f(n/2) + 1$

 $= \Theta(\ \log n)$ $=$ $\log_2 n$ $=f(n/n) + 1 + 1 + ... + 1$

Example-4: Solving Recurrence

- $f(n) = 2 f(n/2) + 1$
- $= 2(2f(n/4)+1)+1$
- $=4f(n/4)+2+1$
- $= 4(2f(n/8)+1)+2+1$
- $= 8f(n/8) + 4 + 2 + 1$
- $= 16 f(n/16) + 8 + 4 + 2 + 1$ $= 8(2f(n/2)+1)+4+2+1$

$$
\ldots \vphantom{\sum_{X_X}^X}
$$

$$
= nf(n/n) + n/2 + ... + 8 + 4 + 2 + 1
$$

=
$$
\sum_{i=0}^{\log n} 2^{i} = \frac{x^{(n+1)} - 1}{x - 1} = \frac{2^{(\log n + 1)} - 1}{2 - 1}
$$

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

28

Example-5: Solving Recurrence

 $= f(n/n) + 2 + 4 + ... + n/8 + n/4 + n/2 + n$... $=f(n/16) + n/8 + n/4 + n/2 + n$ $=f(n/8)+n/4+n/2+n$ $=f(n/4)+n/2+n$ 2 *f(n)=f(n/)+n*

 $= 1 + 2 + 4 + ... + n/8 + n/4 + n/2 + n$

$$
= \sum_{i=0}^{\log n} \frac{n}{2^i} = n \sum_{i=0}^{\log n} \frac{1}{2^i}
$$

this is ^a decreasing series, so

$$
\sum_{i=0}^{\infty} c^i = \frac{1}{1-c} \quad \text{for } |c| < 1
$$

using this we have $c = 1/2$

$$
n\sum_{i=0}^{\infty} \frac{1}{2^i} = n\frac{1}{1-c} = n\frac{1}{1-(1/2)} = n\frac{1}{1/2} = 2n
$$

$$
T(n) \le 2n \in \Theta(n)
$$

29

Selection Problem

- Suppose you have a series of numbers
	- \bullet 5, 7, 2, 10, 8, 15, 21, 37, 41
- How many numbers are smaller than 10
- Rank of a number is the position of number in sorted sequence
- To find the rank of all number we can write $O(n^2)$ algorithm.
- We can sort the numbers

- Sorting takes O(n log n).
- Left to right scanning of array is $O(n)$.

Median

- Median is the middle rank number in a list of number.
- Median is of particular interest in statistics.
- Medians are useful as a measure of central tendency of a set especially when the distribution of values is highly skewed.
- For example, the median income in a community is more meaningful measure than average.

Median (cont'd)

- Suppose 7 households have monthly incomes 5000, 7000, 2000, 10000, 8000, 15000 and 16000
- In sorted order incomes are 2000, 5000, 7000, 8000, 10000, 15000, 16000
- The Median income is 8000; median is element with rank 4: $(7+1)/2=4$
- The average income is 9000

Median (cont'd)

- Suppose the income 16000 goes up to 450,000
- The median is still 8000, but the average goes up to 71000
- Clearly the average is not good representative of the majority of income levels.

Medians & Selection -Sieve Technique

- We will use a variation of divide and conquer strategy called sieve technique.
- In divide & conquer, we break the problem into a small number of smaller subproblems which we solve recursively.
- In Selection Problem we are looking for an item.
- We will divide the problem into subproblems.
- However we will discard those smaller subproblems for which we determine that they do not contain the desired answer (the variation).

- Here is how the sieve technique will be applied to the selection problem
	- We will begin with the given array A[1..n]
	- We will pick an item from the array, called the pivot element which we will denote by x.
	- For now just think of pivot element as a random element of A.

- We partition A into three parts
	- 1.A[q] contains the pivot element x.
	- 2.A[1..q-1] will contain all the elements that are less than x
	- 3.A[q+1..n] will contain all the elements that are greater than x
- Within each sub-array the items may appear in any order.

- The rank of the pivot x is
	- $q p + 1$ in A[p. .r]
- Let rank $x = q p + 1$
- If k = rank_x then the pivot is kth smallest.
- If $k <$ rank x then search A[p..q-1] recursively
- If k > rank_x then search $A[q+1..r]$ recursively. Find element of rank (k-q) because we eliminated q smaller elements in A.

SELECT Algorithm

```
SELECT(array A, int p, int r, int k)
```

```
if (p = r) then return A[p]
```

```
else x \leftarrow CHOOSE_PIVOT(A, p, r)
```

$$
q \leftarrow \text{PARTITION}(A, p, r, x)
$$

rank_x \leftarrow q – p + 1

if $k = rank_x$ then return x

if $k <$ rank_x then return SELECT(A, p, q-1, k) else return SELECT(A, $q+1$, r, $k-q$)

Example: SELECT Algorithm

Select the 6^{th} smallest element of the set $\{5, 9, 2, 6, 4, 1, 3, 7\}$

Analysis of SELECT Algorithm

- For the moment, we will assume that partitioning and pivot both take $\Theta(n)$ time
- How many elements do we eliminate each time?
- If x is the smallest or the largest then we may only succeed in eliminating one element

Pivot is 1

After partitioning

 Ideally, x should have a rank that is neither too large or too small

- Suppose we are able too choose a pivot that causes exactly half of the array to be eliminated in each phase
- This means that we recurse on the remaining n/2 elements
- This leads to the following recurrence

$$
T(n) = \begin{cases} 1 & \text{if } n = 1 \\ T(n/2) + n & \text{otherwise} \end{cases}
$$

If we expand this recurrence we get

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

$$
\leq \sum_{i=0}^{\infty} \frac{n}{2^i}
$$

 \sum ∞ ゠ Ξ $\frac{1}{0}$ 2 1 *i i n*

 Recall the formula for infinite geometric series; for any $|c| < 1$,

$$
\sum_{i=0}^{\infty} c^i = \frac{1}{1-c}
$$

using this we have

$$
n\sum_{i=0}^{\infty} \frac{1}{2^i} = n\frac{1}{1-c} = n\frac{1}{1-(1/2)} = n\frac{1}{1/2} = 2n
$$

$$
T(n) \le 2n \in \Theta(n)
$$

- Lets think about how we ended up with a $\Theta(n)$ algorithm for selection
- Normally a $\Theta(n)$ algorithm would make a single or perhaps a constant number of passes of the data set
- In this algorithm we make a number of passes. In fact it could be as many as log n
- However, because we eliminate a constant fraction of the array with each phase, we get the convergent geometric series in the analysis.