

Introduction to Algorithms - Reading Notes & Selected Solutions

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Solutions to selected exercises

Chapter 2

2.1-1

$A = [31, 41, 59, 26, 41]$

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Algorithm 1 BinaryAddition(A, B, n)

2.1-4

```
carry  $\leftarrow$  0
for  $i \leftarrow 1$  to  $n$ : do
     $C[i] \leftarrow (A[i] + B[i] + C[i]) \bmod 2$ 
     $carry \leftarrow A[i] * B[i]$ 
end for
 $C[i + 1] \leftarrow carry$ 
return  $C$ 
```

- Input: two n -bit numbers A, B .
- Output: the sum of A, B - an $n + 1$ -bit number.

2.2-3 Define X = The number of elements checked in a “brute force” linear search.

Then $X \in \{1 \dots n\}$ and the average number of elements checked in a linear search is exactly:

$$E[X] = \sum_{i=1}^n \frac{1}{n} i = \frac{1}{n} \sum_{i=1}^n i = \frac{n(n-1)}{2n} = \Theta(n)$$

The worst case is where the last element of the array is the one searched for - resulting in an $\Theta(n)$ run time.

2.3-3

$$T(n) = \begin{cases} 2 & n = 2 \\ 2T(\frac{n}{2}) + n & n > 2 \end{cases}$$

Q: Proof by induction that if n is an exact power of two (that is $n = 2^k$ for some constant $k \geq 1$) then $T(n) = n \log n$.

Proof: By induction.

Base case: for $k = 1$ than $T(n) = 2 = 2\log 2 = n\log n$
Assumption: Assume the above holds for all integers up to $k > 1$.
Induction step: We now prove the statement for $n = 2^{k+1}$.
Plugging in to the formula

$$\begin{aligned}
T(n) &= 2T\left(\frac{n}{2}\right) + n = 2T\left(\frac{2^{k+1}}{2}\right) + 2^{k+1} \\
&= 2T(2^k) + 2^{k+1} = 2 * 2^k \log 2^k + 2^{k+1} \\
&= k2^{k+1} + 2^{k+1} = (k+1)2^{k+1} = 2^{k+1} \log 2^{k+1} \\
&= n \log n
\end{aligned}$$

□

Algorithm 2 BinarySearch(A,x)

2.3-5

```

 $l \leftarrow 0$ 
 $r \leftarrow \text{length}(A)$ 
while  $l < r - 1$  do
  if  $x = A[\frac{l+r}{2}]$  then
     $\frac{l+r}{2}$ 
  end if
  if  $x > A[\frac{l+r}{2}]$  then
     $l = A[\frac{l+r}{2}]$ 
  else
     $r = A[\frac{l+r}{2}]$ 
  end if
end while
return -1

```

At each iteration of the while loop the distance between the two pointers - l, r - is halved until the element is found or we return -1. The while loop will terminate once the two pointers are at distance two at which point either the element x is found or the loop will terminate. Thus the distance between the two pointers at each iteration i is precisely $\frac{\text{length}(A)}{2^i} = \frac{n}{2^i}$

At the time of the termination the distance between the two pointers is two, thus -

$$\begin{aligned}
\frac{n}{2^i} &= 2 \\
n &= 2^{i+1} \\
\log(n) &= i + 1 \\
\log(n) - 1 &= i \\
\Theta(\log(n)) &= i
\end{aligned}$$

2-1

2.3-7

Algorithm 3 FindSum(A,x)

```

 $B \leftarrow \text{MergeSort}(A)$ 
 $l \leftarrow 0$ 
 $r \leftarrow \text{length}(B)$ 
while  $l < r$  do
    if  $B[l] + B[r] == x$  then
        return true
    else if  $B[l] + B[r] < x$  then
         $l \leftarrow l + 1$ 
    else
         $r \leftarrow r - 1$ 
    end if
end while
return false

```

- a Given $\frac{n}{k}$ lists each of size k . Applying *InsertionSort* to each list separately yields worst-case runtime of $\Theta(k^2)$. Doing this for all $\frac{n}{k}$ lists yields an $\Theta(\frac{n}{k}k^2) = \Theta(nk)$ runtime algorithm.
- b

Algorithm 4 ModifiedMergeSort(A)

```

Split  $A$  to form  $S = [A_1, \dots, A_{\frac{n}{k}}]$  array of sub-arrays of size  $k$ 
for  $i \leftarrow 1$  to  $\frac{n}{k}$  do
     $A_i \leftarrow \text{InsertionSort}(A_i)$ 
end for
while  $|S| > 1$  do
     $l \leftarrow 1$ 
     $r \leftarrow \text{length}(S)$ 
     $S' \leftarrow \Phi$ 
    while  $l < r$  do
         $S' \leftarrow S' \cup \text{Merge}(A_l, A_r)$ 
         $l \leftarrow l + 1$ 
         $r \leftarrow r - 1$ 
    end while
     $S \leftarrow S'$ 
end while

```

We prove that at each iteration of the outer while loop the size of $|S|$ is $\frac{n}{2^i k}$.

Proof: By induction,

Base $i = 1$: In the first iteration we set l and r to hold the two opposite ends of S , at each iteration we merge two subsets and continue so on until $l = r$ or $l > r$ (depending on the number of subsets) because at each iteration we merged two subsets the number of iterations of the inner loop is precisely $\frac{n}{2k}$.

Step: Assume that the number of subsets in $|S|$ is $\frac{n}{2^i k}$ at iteration i next we prove that at iteration $i + 1$ the above statement holds.

Again, from the same argument for the base case - at each iteration of the inner loop the number of elements decrease by two the number of iterations of the inner loop is $\frac{n}{2^{i+1}k}$ yielding that number of subsets.

The outer loop will terminate once $|S| = 1$, that is -

$$\begin{aligned}\frac{n}{2^i k} &= 1 \\ \frac{n}{k} &= 2^i \\ \log\left(\frac{n}{k}\right) &= i\end{aligned}$$

At each iteration we perform $\frac{n}{2^i k}$ merges each runs in $\Theta(2^i k)$ for a total of $\Theta(n)$, Thus the total running time of the while loop is $\Theta(n \log(\frac{n}{k}))$

All together we get $\Theta(n \log(\frac{n}{k}) + nk)$.

c If one chooses $k = 1$ we get percisly *MergeSort*.

If one chooses $k = n$ we get percisly *InsertionSort*, Thus the choice of k needs to be as close as possible to one. If we choose $k = \Theta(1 - \frac{1}{n}) = \Theta(\frac{n-1}{n})$ which asymptotically is close to one we get -

$$\begin{aligned}T(n) &= \frac{n(n-1)}{n} + n \log\left(\frac{n}{\frac{n-1}{n}}\right) \\ &= n - 1 + n \log\left(\frac{n^2}{n-1}\right) \\ &\approx \Theta(n \log n)\end{aligned}$$

d In practice one can simply use *MergeSort* or if one had to use the modified version, use smaller values of k checking these values “brute force”.

2-4

a The inversions of $[2, 3, 8, 6, 1]$ are

$$(8, 6), (8, 1), (6, 1), (3, 1), (2, 1)$$

b The permutation τ of the set $\{1, \dots, n\}$ with the most inversions is $[n, n-1, n-2, \dots, 1]$ it has $(n-1) + (n-2) + \dots + 1 = \frac{n(n-1)}{2} = \Theta(n^2)$ inversions.

c We prove the following statement (x, y) is in the set of inversions of $S \iff$ its is switched in some iteration of the while loop in the *InsertionsSort* algorithm.

\Leftarrow The pair $(x, y) = (A[i], A[j])$ is switched in some iteration of the *InsertionSort* algorithm, therefore by the loop definition $j = i + 1$ and $y < x$, therefore (x, y) is in the inversions set.

\Rightarrow $(x, y) = (A[i], A[j])$ are in the inversion set of A . we will prove the following Lemma:

Lemma: if $(x, y) = (A[i], A[j])$ are in the inversion set of A than for any integer $i < k \leq j$ the element $A[k]$ is also in the inversion set.

Proof: by induction on the distance between i and j .

Base: $j - i = 1$. Than by defition (x, y) are in the inversion set.

Step: Assume i and j are k elements apart and we prove the statement for i and j at distance $k + 1$. Assume by contradiction that $A[i + k]$ is not an inversion, therfor it is larger than any element that came before it - in particular $A[i]$, For otherwise it would be an inversion. If $A[i + k]$ is larger than $A[j]$ than the pair $(A[i + k], A[j])$ is an inversion for j and i are at distance $k + 1$. If $A[i + k]$ is smaller than $A[j]$ than the pair $(A[i + k], A[i])$ is an inversion since $A[i] > A[j] > A[i + k]$ by the assumption the $(A[i], A[j]) = (x, y)$ is in the inversion set. \square

By the Lemma any element between $A[i]$ and $A[j]$ are in the inversion set, that means that in the inner loop of *InsertionSort* all of those elements will be switched in the inner loop, including (x, y) .

Conclusion: The number of elements in the inversion set of A is precisely the number of iterations of the inner loop of *InsertionSort*. In other words, the run time of *InsertionSort* is $\Theta(|S|)$.

Chapter 5

5.1-2

Algorithm 5 Random(a,b)

```

if  $a = b$  then
     $a$ 
end if
while  $a < b$  do
    if  $\text{Random}(0, 1) > 0$  then
         $\text{Random}(a, \frac{a+b}{2})$ 
    else
         $\text{Random}(\frac{a+b}{2}, b)$ 
    end if
end while

```

The runtime of the above algorithm is $O(\log(\frac{b-a}{2}))$.

- 6.1-1 The maximum number of elements in a heap of height h is $2^0 + 2^1 + \dots + 2^h = \sum_{i=1}^h 2^i = \frac{2^{h+1}-1}{2-1} = 2^{h+1} - 1$. The minimum number of elements occurs when there is precisely one leaf node (i.e. the bottom level of the binary-tree is empty but one element), meaning:

$$2^0 + 2^1 + \dots + 2^{h-1} + 2 = \sum_{i=0}^{h-1} 2^i + 1 = \frac{2^h - 1}{2 - 1} + 1 = 2^h$$

- 6.1-2 Proof by Induction:

Base: $n = 1$, A single node heap is at height $0 = \log(1) = \log(n)$

Step: We assume the statement is correct for $n = k - 1$ and prove for a heap of size $n = k$. Consider the last element of the heap $A[k]$, by the definition of the heap, its parent is the element $A[\frac{k}{2}]$. By the induction step the heap $A[1 \dots \frac{k}{2}]$ is of height $\log(\frac{k}{2}) = \log(k) - \log(2) = \log(k) - 1$. Thus, the height of the heap $A[1 \dots k]$ has one more layer than $A[1 \dots \frac{k}{2}]$, that is $\log(k)$.

□

- 6.1-7 We prove the counter-positive statement, that is, if a node is indexed by $i \in \{1 \dots \frac{n}{2}\}$ in the array representation of the heap then it is **not** a leaf node.

By contradiction, assume it was indexed by $i \in \{\frac{n}{2} + 1, \dots, n\}$ then, either one of his children had to be at some index k such that

$$k \geq 2i \geq 2(\frac{n}{2} + 1) \geq n$$

Therefore it exceeds the size of the heap - contradiction.

Conclusion: a node is indexed by $i \in \{\frac{n}{2} + 1, \dots, n\} \iff$ it is a leaf node in the heap.

Chapter 6

6-2

- a One can represent a d -ary heap using the following structure: for every non leaf node indexed by i in the array A , its children are indexed in the array at positions $Children(i) = A[i*d+1, \dots, i*d+d]$.
- b The height of a d -ary heap is $O(\log_d n)$.
- c First we need to modify *MAX-HEAPIFY* for a d -ary heap:

Once the new heapify is defined we can use the same procedure *Extract-Max* as defined in the book except use our d -ary *MAX-HEAPIFY* in line 6.

Algorithm 6 d-ary MAX-HEAPIFY(A, i)

$Children \leftarrow Children(i)$
 $largest \leftarrow i$
for j in $1 \dots d$: **do**
 if $A[i] > A[largest]$ **then**
 $largest \leftarrow i$
 end if
end for
if $i \neq largest$ **then**
 $A[i] \leftrightarrow A[largest]$
 d-ary MAX-HEAPIFY($A, largest$)
end if
