

Homework 1 Theoretical Solutions (150 points)**1. Solution to Problem 1**

Algorithm: we first find the value of n in $O(\log n)$ and use Binary search to find the index in $O(\log n)$ time. To find the value in $O(\log n)$ time, we first set $\text{left}=0$ and $\text{right}=1$, compare from $A[0]$ to $A[\text{right}]$ and the input integer, if it is greater than right index element then copy right index in left index and double the right index ; if it is smaller, then apply binary search on left and right indices found.

The worst run time will exam k right indices, where $A[2^k]=\infty$, to leave the comparison, thus $2^{k-1} \leq n \leq 2^k$, thus run time is $O(\log 2^k)$, the binary search will search k times and thus run time is $O(\log 2^k)$, thus total run time is $O(\log 2^k)$. Since $2^k \leq 2n$, thus $O(\log_2(2n)) = O(\log n)$ correctness is the same as binary search; $O(\log n)$ run time

Algorithm 1 Algorithm binary search

Input 1. a sorted array A of n integers. 2. integer x

Output The index of x if exist; ∞ if not.

```

procedure Search( $A, x$ )
     $\text{left}=0, \text{right}=1, \text{value}=A[0]$ 
    if  $x > \text{value}$  then
         $\text{left} = \text{right}, \text{right} = 2 * \text{right}, \text{value} = A[\text{right}]$ 
    end if
    return BinarySearch( $A, x, \text{left}, \text{right}$ )
end procedure

```

2. Solution to Problem 2

(a) a time: $O(n) * O(n - 1) = O(n^2)$

correctness: iteratively go through all $n > j > i$ for each i and compare, if $A[i]$ is greater, count pair, to get total number of pairs of disorder

Algorithm 2 Algorithm brute force

Input a sequence of n distinct numbers

Output number of pairs of disorder

```
procedure Bruteforce( $A$ )  
    pair=0, n=len( $A$ )  
    for  $i \in [1, n]$  do  
        while  $j \in [i+1, n]$  do  
            if  $A[i] > A[j]$  then  
                pair+=1  
            end if  
        end while  
    end for  
    return pair  
end procedure
```

- (b) Algorithm: Mergesort(A, l, r) recursively divide the array into sub-problems and count the pairs of disorder. Then use Merge to combine the sub-arrays and count the total number of pairs.
 run time of mergesort: $O(n \log n)$; Merge is $O(n)$ time
 correctness of Mergesort and Merge is proven in class

Algorithm 3 Algorithm MergeSort Count

input: a sequence of n distinct numbers

output: number of pairs of disorder

procedure MergeSort(A, l, r)

 pair=0

if $l = r$ **then return** 0

else if $l < r$ **then**

 mid=($l+r$)/2

 pair += MERGESORT(A, l, mid)

 pair += MERGESORT($A, \text{mid}+1, r$)

 pair += MERGE(A, l, mid, r)

end if

return pair

end procedure

input: List A is comprised of two sorted lists

output: number of pairs disorder

procedure Merge(A, l, mid, r)

 pair=0, $i=l$, $j=\text{mid}+1$, $k=l$, output=list for output with length of A

while $i \leq \text{mid}$ & $j \leq r$ **do**

if $A[i] \leq A[j]$ **then**

 output[k]= $A[i]$, $i+=1$, $k+=1$

else

 output[k] = $A[j]$, pair+=($\text{mid}-i+1$), $k+=1$, $j+=1$

 ▷ $A[j]$ is less than every number from index i to mid

end if

end while

while $i \leq \text{mid}$ **do** copy remaining of left to the end of output list

end while

while $j \leq r$ **do** copy remaining of right to the end of output list

end while

return pair

end procedure

3. (25 points) In the table below, indicate the relationship between functions f and g for each pair (f, g) by writing “yes” or “no” in each box. For example, if $f = O(g)$ then write “yes” in the first box. Here $\log^b x = (\log_2 x)^b$.

f	g	O	o	Ω	ω	Θ
$6n \log^2 n$	$n^2 \log n$	yes	yes	no	no	no
$\sqrt{\log n}$	$(\log \log n)^3$	yes	yes	no	no	no
$10n \log n$	$n \log(10n^2)$	yes	no	yes	yes	yes
$n^{4/5}$	$\sqrt{n} \log n$	no	no	yes	yes	no
$5\sqrt{n} + \log^2 n$	$3\sqrt{n}$	yes	no	yes	no	yes
$\frac{3^n}{n^3}$	$n^3 2^n$	no	no	yes	yes	no
$\sqrt{n} 2^n$	$2^{n/2 + \log n}$	no	no	yes	yes	no
$n \log(2n)$	$\frac{n^2}{\log n}$	yes	yes	no	no	no
$n!$	n^n	yes	yes	no	no	no
$\log n!$	$\log(n^{n/2})$	yes	no	yes	no	yes

4. (a) **Proof :**

- i. Base case:

$$F_6 = 8 \geq 2^3$$

$$F_7 = F_6 + F_5 = 13 \geq 2^{7/2}$$

- ii. Inductive hypothesis:

suppose for all $n \geq 6, F_n \geq 2^{n/2}$ holds.

- iii. inductive step:

For $n > 6$, the following holds

$$F_{n+1} = F_n + F_{n-1} \geq 2^{n/2} + 2^{(n-1)/2} = 2^{n/2} * (1 + \frac{1}{\sqrt{2}}) > 2^{n/2} \sqrt{2} = 2^{(n+1)/2}$$

Thus, proved.

- (b) i) Give an algorithm that computes F_n based on the recursive definition above. Develop a recurrence for the running time of your algorithm and give an asymptotic lower bound for it.

Run time:

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

$$\begin{aligned}
&= 3T(n-3) + 2T(n-4) + 4c \\
&\geq 2T(n-3) + 2T(n-4) + 4c \\
&= 4T(n-4) + 2T(n-5) + 6c \\
&\dots \\
&\geq 2^k T(n-2k) + (2^{k+1} - 2)c \\
&\text{let } n-2k = 0, \text{ thus } T(n) = \Omega(2^{n/2})
\end{aligned}$$

Correctness:

Base: if $n=0$ $n=1$, $\text{Recurfib}(n) = F_n = n$ is correct

Hypothesis: for $n > 1$, $\text{Recurfib}(n)$ correctly computes F_n

Inductive Step: By definition of the algorithm

$$\text{Recurfib}(n+1) = \text{Recurfib}(n) + \text{Recurfib}(n-1)$$

$$= F_n + F_{n-1} = F_{n+1}$$

Thus the algorithm is correct.

Algorithm 4 Algorithm Fib Recursion

Input An integer n

Output n^{th} Fibonacci number

procedure $\text{Recurfib}(n)$

if $n \leq 1$ **then return** n

else

return $\text{Recurfib}(n-1) + \text{Recurfib}(n-2)$

end if

end procedure

- ii) 8 points) Give a non-recursive algorithm that asymptotically performs fewer additions than the recursive algorithm. Discuss the running time of the new algorithm. Description: First set the base cases, to find the n th fib number, iterate $n-2$ times: next fib is sum of the last two which $a+b$, move $a=b$, $b=c$ to start the next iteration until done.

Run time: since there is only one for loop: $O(n)$

Correctness:

Base: fib(n) is correct for $n=0$ and $n=1$

Hypothesis: fib(n) return Fibonacci number correctly for $n > 1$

Inductive step:

Since fib(n) and fib($n-1$) return correct F_n and F_{n-1}

We will show fib($n+1$) returns F_{n+1}

from the algorithm, at the end of $n-3$ iteration, $b=\text{fib}(n-1)=F_{n-1}$; at the end of $n-2$ iteration, $a=\text{fib}(n-1)$, $b=\text{fib}(n)=F_n$; at the end of $n-1$ iteration $c=\text{fib}(n)+\text{fib}(n-1)=F_n+F_{n-1}=F_{n+1}$, return $b=c=F_{n+1}$ correctly

Algorithm 5 Algorithm Fib Non-Recursion

Input An integer n

Output n^{th} Fibonacci number

procedure fib(n)

$a=0, b=1$

if $n \leq 1$ **then return** n

else

for $i \in [1, n-2]$ **do**

$c=a+b, a=b, b=c$

end for

end if

return b

end procedure

iii) Description:

To get the nth fib number, we do n-1 power of matrix M and multiply by matrix $\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$ and get the first element in the matrix. Improve (n-1) power by using (n-1)/2 power of matrix multiply by itself; If n is even, we get (n-1) power we need; If n is odd, multiply by M and get n-1 power we need.

Run time of Power(M,n): $T(n) = T(n/2) + c = O(\log n)$

Run time of fib(n): $T(n) = O(\log n) + O(1) = O(\log n)$

Correctness of Power(M,n):

We know Multiply(A,B) is correct because it is just multiplication and addition. We need to prove the algorithm returns correct Matrix power of n

Base: n=1, Power(M,n) is correct

Hypo: assume for any $n > 1$, Power(M,n) gives correct matrix of n power to M.

Inductive step:

if n is even, since Power(M,n/2) is correct, then multiply by itself gives correct n power.

if n is odd, since Power(M,n/2) is correct, then multiply by itself and K gives correct power $2*(n/2)+1 = n$. proved.

Correctness of fib(n):

We know Power(M,n) is correct.

Base: fib(0) and fib(1) = $M^0 * F_{base}[0][0]$ is correct;

Inductive hypothesis: assume fib(n) correctly return F_n

Inductive Step:

For n+1, $Power(M, n) * F_{base} = M * Power(M, n-1) * F_{10} = M * [F_n \ F_{n-1}]^T = [F_n + F_{n-1} \ F_n]^T$ which will return the first element of the matrix: $F_n + F_{n-1} = F_{n+1}$, proved.

Algorithm is located the next page

Algorithm 6 Algorithm Fib Matrix

Input An integer n

Output n^{th} Fibonacci number

procedure fib(n)

$$M = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix}$$

$$F_{base} = [1 \ 0]$$

if $n = 0$ **then return** 0

else

return Power($M, n-1$) * $F_{base}[0][0]$

▷ Return the first element of the matrix

end if

end procedure

procedure Power(M, n)

$$K = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix}$$

if $n \leq 1$ **then return** M

else

$M = \text{Power}(M, n/2)$, $M = \text{multiply}(M, M)$

if n is odd **then**

$M = \text{Multiply}(K, M)$

end if

end if

end procedure

procedure Multiply(A, B)

$$x = (A[0][0] * B[0][0] + A[0][1] * B[1][0])$$

$$y = (A[0][0] * B[0][1] + A[0][1] * B[1][1])$$

$$z = (A[1][0] * B[0][0] + A[1][1] * B[1][0])$$

$$w = (A[1][0] * B[0][1] + A[1][1] * B[1][1])$$

$$A[0][0] = x$$

$$A[0][1] = y$$

$$A[1][0] = z$$

$$A[1][1] = w$$

return A

end procedure

- iv) This algorithm is the same as the last part except the matrix M and the F_{base} are different. Run time $O(\log n)$ and correctness can both be proved similarly to the previous question.

Algorithm 7 Algorithm Fib Plus Number

Input An integer n

Output n^{th} Fibonacci Plus number

procedure $p(n)$

$$M = \begin{bmatrix} 1 & 3 & 2 & 0 \\ 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 1 \\ 1 & 0 & 0 & 1 \end{bmatrix}$$

$$F_{base} = [p_1 \ p_0 \ 2 \ 1]$$

if $n = 0$ **then return** a

else

return $\text{Power}(M, n-1) * F_{base}^T[0][0]$

end if

end procedure

procedure $\text{Power}(M, n)$

$$K = \begin{bmatrix} 1 & 3 & 2 & 0 \\ 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 1 \\ 1 & 0 & 0 & 1 \end{bmatrix}$$

if $n \leq 1$ **then return** M

else

$M = \text{Power}(M, n/2), M = \text{Multiply}(M, M)$

if n is odd **then**

$M = \text{Multiply}(K, M)$

end if

end if

end procedure

procedure $\text{Multiply}(A, B)$

$A[i][j] = \sum_{k=1}^4 (A[i][k] * B[k][j])$ **return** A

5. (a) Run time is $O(1)+O(n-1)*O(n-2)*O(1)=O(n^2)$ since there are two loops

Description and Correctness: Initialize min to be very large, then we iteratively calculate the euclidean distance of p_i and all p_j for $j \in [i+1, n]$, for every $i \in [1, n-1]$ and obtain the min by comparing with the previous min, the pair with smallest distance is the closest.

Algorithm 8 BruteForce Closest Pair

Input a set of n points, (n is a power of 2) in the plane $P = \{p_1 = (x_1, y_1), p_2 = (x_2, y_2), \dots, p_n = (x_n, y_n)\}$.

Output the pair (p_i, p_j) with $p_i \neq p_j$ for which the euclidean distance between p_i and p_j is minimized.

```
procedure BF( $P$ )
    min =  $\infty$ 
    for  $i$  in  $[1, n-1]$  do
        for  $j$  in  $[i+1, n]$  do
            if euclidean distance of  $p[i]$  and  $p[j] < \text{min}$  then
                min = the euclidean distance, pair =  $(p[i], p[j])$ 
            end if
        end for
    end for
    return pair
end procedure
```

- (b) i. Let $T(n)$ be the run time of the algorithm.
1. Find a value x for which exactly half the points have $x_i < x$ and half have $x_i > x$. On this basis, split the points in two groups, L and R. We need to sort by x coordinates first, which is $O(n \log n)$ using mergesort
 2. Recursively find the closest pair, which takes $2 * T(n/2)$
 3. Discard all points with $x_i < x - d$ or $x_i > x + d$ and sort the remaining points by y coordinate. Finding the points to discard and Mergesort takes $O(n) + O(n \log n)$
 4. Now go through the sorted list and for each point compute its distance to the seven subsequent points in the list. $O(n) * 7 = O(n)$
- The answer is whichever has the smallest euclidean distance.
- Thus, Recurrence run time = $T(n) = 2 * T(n/2) + cn \log n + cn = 2T(n/2) + cn \log n$. Solving this in the next part.

Algorithm 9 Divide and Conquer Closest Pair

Input two set of n points $\{xp\}, \{yp\}$.

Output the distance and the pair (p_i, p_j) with $p_i \neq p_j$ for which the euclidean distance between p_i and p_j is minimized.

```

procedure Cloest( $xp, yp$ )
    if less than 3 points then use brute force
    else
         $xp = \text{Mergesort}(xp)$ 
         $xm = xp[n/2]$ 
         $xl =$  points that has  $x$  coordinate less than  $xm$ ,  $yl$  is corresponding  $y$  coordinate
         $xr =$  points that has  $x$  coordinate greater than  $xm$ ,  $yr$  is corresponding  $y$  coordinate
         $dl, (pl, ql) = \text{Cloest}(xl, yl)$ 
         $dr, (pr, qr) = \text{Cloest}(xr, yr)$ 
    end if
     $dmin = \min(dl, dr)$ , minpair is pair with min distance
    remove  $(x, y)$  such that  $|xm - x| > dmin$ 
     $p_{sorted} = \text{Mergesort}$  the remaining points by  $y$  coordinate
    for  $i$  in length of  $p_{sorted}$  do
         $k = i + 1$ 
        while  $k \leq \text{len}(p_{sorted})$  do
            if euclidean distance  $(p_{sorted}[i], p_{sorted}[k]) < dmin$  then
                 $dmin =$  euclidean distance of  $p_{sorted}[i]$  and  $p_{sorted}[k]$ 
                 $pair = (p_{sorted}[i], p_{sorted}[k])$ 
            end if
             $k += 1$ 
        end while
    end for
    return pair,  $dmin$ 
end procedure

```

- ii. Let $T(n)$ be the run time of the algorithm.

1. Find a value x for which exactly half the points have $x_i < x$ and half have

$x_i > x$. On this basis, split the points in two groups, L and R. We need to sort by x coordinates first, which is $O(n \log n)$ using mergesort

2. Recursively find the closest pair, which takes $2T(n/2)$

3. Discard all points with $x_i < x - d$ or $x_i > x + d$ and sort the remaining points by y coordinate. Finding the points to discard and Mergesort takes $O(n) + O(n \log n)$

4. Now go through the sorted list and for each point compute its distance to the seven subsequent points in the list. $O(n) \cdot 7 = O(n)$

The answer is whichever has the smallest euclidean distance.

Thus, Recurrence run time = $T(n) = 2T(n/2) + c n \log n + c n = 2T(n/2) + c n \log n$. Solving this in the next part.

$$T(n) = 2T(n/2) + c n \log n + c n = 2T(n/2) + c n \log n$$

$$= 2(2T(n/4) + (n/2) \log(n/2)) + n \log n$$

$$= 4T(n/4) + n \log(n/2) + n \log n$$

...

$$= 2^k T(n/2^k) + n(\log n + \dots + \log(n/2^k))$$

let $n = 2^k$, then $k = \log n$, $T(1) = 1$ thus:

$$T(n) = 2^k T(n/2^k) + n(\log n + \dots + \log(n/2^k)) < n + n k \log(n) = n + n \log^2(n) = O(n \log^2(n))$$