

1.

a)

For $j \leftarrow 1$ to $n - 1$: # main loop that iterates $n - 1$ times:

For $i \leftarrow n$ to 2 : # visits the elements of A in decreasing order of indices:

If $A[i] > A[i - 1]$ then # compares $A[i]$ with $A[i - 1]$:

swap($A[i], A[i - 1]$) # if the statement is true, do the swap

b)

In such algorithm it guaranteed that $A[i - 1]$ is larger than $A[i]$ after first i passed

for base case: $n = 2$

if $A[2] > A[1]$ then do the swap otherwise the array already sorted so it is correct

Assuming that algorithm sorts the array size of $n - 1 \geq 2$ correctly, such that $[A[2] \text{ to } A[n]]$ is in descending order

first add a largest number in A , index as 1, with above induction, this number $A[1]$ would be larger than $A[2]$ as guaranteed that $A[i - 1]$ is larger than $A[i]$ after first i passed

As now $[A[2] \text{ to } A[n]]$ is sorted in descending order (assumed)

$[A[1] \text{ to } A[n]]$ would also be sorted in descending order

if the adding number α is not $> A[2]$ and still adding in index 1 as $A[1]$. and for a number k , $k \leq n$ such that $A[k - 1] > \alpha \geq A[k]$, as guaranteed above, after $k - 1$ times iterations each time of $A[i = k \text{ to } 2] > \alpha$

such that α would move to $A[k - 1]$, with such induction $A[1] \dots A[n]$ would be sorted in decreasing order so the algorithm is correct as $n - 1$ case correct implied n case correct

c)

As this algorithm would run $n - 1$ times iterations and in each iterations it will do $n - 1$ times comparisons, so in total the worst case running time would be $(n - 1)(n - 1) = n^2 - 2n + 1$

so the bound on the worst - case running time would be $O(n^2)$

2.

a)

let B be a list for storing the answer

function permutation(temporary array α):

 if length of α equals to length of A then

 add α in B

 return # terminate the function call

 For $i \leftarrow 1$ to n : # iterate n times and add element i to temporary array α

 if $A[i]$ is not in α then

$A[i]$ append in α

 permutation(α) # recursively call the function with added element

permutation(empty array $[\]$) # initial call of the function

the return would be B as a array storing all the permutation array

b)

Recurrence for the running time of recursive calling the function in(a)

$T(n) = nT(n - 1)$ with boundary condition $T(1) = 1$

by solving the recurrence $T(n - 1) = (n - 1)T(n - 2)$

$T(n) = n(n - 1)T(n - 2)$

In general case $T(n) = n!$ such that $T(n) = n! = O(n!) = \Omega(n!)$ such that $T(n) = \Theta(n!)$

3.

a) $\frac{n^2}{\log(n)} = O(n^2)$

b) $2^{\log(n)} = O(n^3)$

c) let $k = \log(n)$, $\log(k) = O(k)$, so $\log(\log(n)) = O(\log(n))$

d) as $4^{\log_2 n} = n^{\log_2 4} = n^2$ such that $n^2 = \Theta(n^2)$ so $n^2 = \Theta(4^{\log_2 n})$, $4^{\log_2 n} = \Theta(n^2)$

e) $(\log n)^2 = O(n^{1/3})$

4. Master theorem

$$T(n) = aT\left(\frac{n}{b}\right) + n^d \quad \text{if } d < \log_b a, \quad \Theta(n^{\log_b a})$$

$$T(n) = n^{\log_b a} + n^d \left(\frac{\left(\frac{a}{b^d}\right)^{\log_b n} - 1}{\left(\frac{a}{b^d}\right) - 1} \right) \quad \text{if } d = \log_b a, \quad \Theta(n^d \log n)$$

$$T(n) = n^{\log_b a} + n^d \left(\frac{\left(\frac{a}{b^d}\right)^{\log_b n} - 1}{\left(\frac{a}{b^d}\right) - 1} \right) \quad \text{if } d > \log_b a, \quad \Theta(n^d)$$

a) $a=4$ $b=2$ $d=1$

$$1 < \log_2 4 \quad \text{so } d < \log_b a$$

So the upper bound be $\Theta(n^{\log_2 4}) = \Theta(n^2)$

b) $a=4$ $b=2$ $d=3$

$$3 > \log_2 4 \quad \text{so } d > \log_b a$$

So the upper bound would be $\Theta(n^3)$

$$c) \quad a=1 \quad b=3 \quad d=\frac{1}{2}$$

$$\frac{1}{2} > \log_3 1 \quad \text{so} \quad d > \log_b a$$

So the upper bound be $O(n^{\frac{1}{2}})$

d)

$$T(n) \leq 5T\left(\frac{n}{4}\right) + n \log n$$

$$T\left(\frac{n}{4}\right) \leq 5T\left(\frac{n}{4^2}\right) + \log\left(\frac{n}{4}\right) \left(\frac{n}{4}\right)$$

$$T(n) \leq 5 \left(5T\left(\frac{n}{4^2}\right) + \left(\frac{n}{4}\right) \log\left(\frac{n}{4}\right) \right) + n \log n$$

$$= 5^2 \left(T\left(\frac{n}{4^2}\right) + \left(\frac{n}{4}\right) \log\left(\frac{n}{4}\right) \right) + n \log n$$

$$\vdots = 5^2 T\left(\frac{n}{4^2}\right) + 5 \left(\frac{n}{4} \right) \log\left(\frac{n}{4}\right) + n \log n$$

In general case. $as \quad 4^i = n$
 $i = \log_4 n$

$$T(n) \leq 5^i T\left(\frac{n}{4^i}\right) + n \sum_{j=0}^{i-1} \left(\frac{5^j}{4^j} \right) \log\left(\frac{n}{4^j}\right)$$

As i goes to $\log_4 n$

$$T(n) = n^{\log_4 5} + \underbrace{\sum_{j=0}^{\log_4 n - 1} \left(\frac{5}{4} \right)^j}_{\text{constant}} \sum_{j=0}^{\log_4 n - 1} \log\left(\frac{n}{4^j}\right) n$$



$$\log\left(n \sum_{j=0}^{\log_4 n - 1} \left(\frac{1}{4}\right)^j\right) n$$

$$\frac{1 - \left(\frac{1}{4}\right)^{\log_4 n}}{1 - \frac{1}{4}} = \frac{4 - \frac{4}{n}}{3}$$

simplify to $\Rightarrow T(n) = n^{\log_4 5} + (\text{constant})(n)(\log n)$

As $n^{\log_4 5}$ grow faster than $n \log n$

so $T(n) = O(n^{\log_4 5})$

e)

$$T(n^{\frac{1}{2}}) = T(\lfloor n^{\frac{1}{2}} \rfloor) + 1$$

$$T(n) = (T(\lfloor n^{\frac{1}{2}} \rfloor) + 1) + 1$$

For general case it is inducted to be

$T(n) = (T(\lfloor n^{\frac{1}{2^i}} \rfloor) + i)$ as i to be the minimal of levels to base case $T(1)$

as $\lfloor \cdot \rfloor$ is a floor function while

$1 \leq n^{\frac{1}{2^i}} < 2$ it would be $= 1$

$$\text{So } n^{\frac{1}{2^i}} < 2 \rightarrow \frac{1}{2^i} < \frac{\log 2}{\log n}$$

$$\rightarrow \log_2 n < 2^i \quad \log_2 \log_2 n < i \Rightarrow \lceil \log_2 \log_2 n \rceil$$

as inducted, $\log_2 \log_2 n$ be the minimal levels

$$\text{to base } T(1) \quad T(n) = 1 + \log_2 \log_2 n$$

$$\text{such that } T(n) = O(\log \log n)$$

5.

use a divide and conquer algorithm

#binary search

$l \leftarrow 1$

$r \leftarrow n$

While $l < r$:

$m = (l + r) // 2$ *#divide into 2 subarray and floor it if $(l + r)$ is not divisible by 2*

If r is $l + 1$:

return $A[r]$

if $A[m] > A[m + 1]$: #check if the unique minimum stay in the right sides or left

#if $A[m] > A[m + 1]$ it means the unique minimum would stay in right side

$l = m$ *#then subtract into a half size array in this case as right side*

else:

$r = m$ *#otherwise left side*

For $n = 3$, is a base case, by $[n_1, n_2, n_3]$ such that by definition n_2 appear to be the unique minimum then $A[m] = n_2$, $m = 2$, $A[m] < A[m + 1]$ so $r = m$, $A[r] = n_2$, $A[l]$ still be n_1 and $l = 1$

so $r = l + 1$ return $A[r] = n_2$ to be the unique minimum

For $n > 3$

Assuming the algorithm can correctly find the unique minimal γ of index r_γ in a n size array A

by appending a new element α , there will be 2 cases

1. $\alpha > \gamma$

By the definition of convex function, α would live in $A[1]$ to $A[r_\gamma - 1]$ or $A[r_\gamma + 1]$ to $A[n]$

with the algorithm running, either the α would get eliminated in the progress of division to subarray

or if γ next to α , take m as r_γ , $A[r_\gamma] < A[r_\gamma + 1]$ so $r = r_\gamma$ as $l = r_\alpha$, $r_\gamma = r_\alpha + 1$, return $A[r_\gamma] = \gamma$

such that γ would still be the unique minimum

2. $\alpha < \gamma$

By the definition of convex function, α would live in next to γ , take $m = r_\gamma$ then $A[r_\gamma] > A[r_\gamma + 1] = A[r_\alpha]$

as now $l = r_\gamma$ with $r_\alpha = r_\gamma + 1$, the algorithm would return $A[r_\alpha]$ which to be unique minimum α

so in two cases inducted that the algorithm is correct as true in case n implies case $n + 1$ is true

As in each loop do one comparison of $A[m]$ and $A[m + 1]$ then divided in $A[1 \dots m]$ and $A[m + 1 \dots n]$

so $T(n) = T(n/2) + 1$ by master theorem $a = 1$, $b = 2$, $d = 0$

$0 = \log_2 1$

$T(n) = O(\log(n))$

6.

The algorithm would base on divide every polynomial function into a quadratic function and a constant term such that we can use the property $(ax + c)(bx + d) = (ab)x^2 + (ad + bc)x + (cd)$ with further induction $(ad + cb) = (a + c)(b + d) - (ab) - (cd)$ in such case we only need to compute (ab) , (bc) and $(a + c)(b + d)$ to obtain the coefficient of x^2 , x^1 and x^0

For $p(x)$ and $q(x)$ are polynomial function as it can divide to

$$p = (p_0)x + (p_1)x^0$$

$$q = (q_0)x + (q_1)x^0$$

$$pq = (p_0q_0)x^2 + (p_0q_1 + q_0p_1)x + (p_1q_1)x^0$$

as inducted above

$$(p_0q_1 + q_0p_1) = (p_0 + p_1)(q_0 + q_1) - (p_0q_0) - (p_1q_1)$$

$$pq = (p_0q_0)x^2 + ((p_0 + p_1)(q_0 + q_1) - (p_0q_0) - (p_1q_1))x + (p_1q_1)x^0$$

A simple version code would be like

function $A(p, q, n)$:

$$P_0Q_0 = A((p[n]x^{n-1} + p[n-1]x^{n-2} + p[n-2]x^{n-3} + \dots), (q[n]x^{n-1} + q[n-1]x^{n-2} + q[n-2]x^{n-3} + \dots))$$

$$P_1Q_1 = A((p[0]), (q[0]))$$

$$(P_0 + P_1)(Q_0 + Q_1)$$

$$= A((p[n]x^{n-1} + p[n-1]x^{n-2} + p[n-2]x^{n-3} + \dots + p[0]), (q[n]x^{n-1} + q[n-1]x^{n-2} + q[n-2]x^{n-3} + \dots + q[0]))$$

$$\text{return } (P_0Q_0)x^2 + ((P_0 + P_1)(Q_0 + Q_1) - (P_0Q_0) - (P_1Q_1))x + (P_1Q_1)x^0$$

by recursively calling the function with input p_0q_0 , $(p_0 + p_1)(q_0 + q_1)$ and (p_1q_1)

by merging back and return with above equation, we can get all the coefficient up to $2n$ terms

By assuming n is a power of 2, $n = 2^k$

there are 3 function calling divisions of each size of 2 also merging with above size n

so $T(n) = 3T(n/2) + O(n)$

by master theorem, $a = 3, b = 2, d = 1, T(n) = \Theta(n^{\log_b a})$ if $d < \log_b a$

$1 < \log_2 3$, so $T(n) = \Theta(n^{\log_2 3}) = O(n^{\log_2 3})$ in such case less than $\Theta(n^2)$ and meet $o(n^2)$

For $n = 0$, is the base case returning $P[0]Q[0]$

For $n = 1$, also true with returning $(P[0]Q[0])x^2 + (P[0]Q[1] + P[1]Q[0])x + (P[1]Q[1])x^0$

For $n > 1$ assuming the algorithm would return correct answer with recursively divide the polynomial into

$P[n]x^n + p[n-1]x^{n-1} + p[n-2]x^{n-2} + \dots + P[0]x^0$

$(p[n]x^{n-1} + p[n-1]x^{n-2} + p[n-2]x^{n-3} + \dots)x + p[0]x^0$

$P_0 = (p[n]x^{n-1} + p[n-1]x^{n-2} + p[n-2]x^{n-3} + \dots)x, P_1 = p[0]x^0$

as doing the same thing to Q

we can get $P_0Q_0 = (p[n]x^{n-1} + p[n-1]x^{n-2} + p[n-2]x^{n-3} + \dots)(q[n]x^{n-1} + q[n-1]x^{n-2} + q[n-2]x^{n-3} + \dots)$

$P_1Q_1 = (p[0])(q[0])$

$(P_0 + P_1)(Q_0 + Q_1)$

$= (p[n]x^{n-1} + p[n-1]x^{n-2} + p[n-2]x^{n-3} + \dots + p[0])(q[n]x^{n-1} + q[n-1]x^{n-2} + q[n-2]x^{n-3} + \dots + q[0])$

with above assumption is correct we can see for $n + 1$ case

$P_0Q_0 = (p[n+1]x^n + p[n]x^{n-1} + p[n-1]x^{n-2} + \dots)(q[n+1]x^n + q[n]x^{n-1} + q[n-1]x^{n-2} + \dots)$

as let $k = n + 1$

we get $(p[k]x^{k-1} + p[k-1]x^{k-2} + p[k-2]x^{k-3} + \dots)(q[k]x^{k-1} + q[k-1]x^{k-2} + q[k-2]x^{k-3} + \dots)$

$(P_0 + P_1)(Q_0 + Q_1) =$

$(p[k]x^{k-1} + p[k-1]x^{k-2} + p[k-2]x^{k-3} + \dots + p[0])(q[k]x^{k-1} + q[k-1]x^{k-2} + q[k-2]x^{k-3} + \dots + q[0])$

same as the n term so it is correct

P_1Q_1 still be $(p[0])(q[0])$

such that if n case is correct indicate $n + 1$ case is also correct so the algorithm works correctly