

DSA Midterm Exam 2020

答案仅供参考!

Q1

Which of the following is (are) correct?

☒ n^2 in Theta (n^2)

☐ n^3 in $O(n^2)$

☒ n^2 in $O(n^3)$

☐ n^3 in Theta (n^2)

☐ n^2 in Theta (n^3)

☒ n^2 in $O(n^2)$

Q2

What is the minimal number of leaves that a decision tree for some sorting algorithm for an input of size n has?

☐ $2^{(n-1)}$

☒ $n!$

☐ $n\log(n)$

☐ 2^n

Q3

What is the result of adding the key 10 to the max-heap [7,6,5,4,3,2,1,0,0,0] with initially heap size 7?

We apply the adding-operation on the fly.

(But see attached pseudo-code)

Algorithm `heapIncreaseKey(H, i, k):`
 if $k < H[i]$ **then**
 return error
 $H[i] := k$
 while $i > 1$ **and** $H[\text{parent}(i)] < H[i]$ **do**
 `swap($H[\text{parent}(i)], H[i]$)`
 $i := \text{parent}(i)$

Algorithm `heapInsert(H, k):`
 $H.\text{heap-size} := H.\text{heap-size} + 1$
 $H[H.\text{heap-size}] := -\infty$
 `HeapIncreaseKey($H, H.\text{heap-size}, k$)`

- ☐ [10,7,6,5,2,3,4,1,0,0] with heapsize 10.
- ☒ [10,7,5,6,3,2,1,4,0,0] with heapsize 8.
- ☐ [7,5,6,1,2,3,4,10,0,0] with heapsize 8.
- ☐ [10,7,6,5,4,3,2,1,0,0] with heapsize 8.

Q4

In heapsort we use the subroutine `MaxHeapify`(see the attached pseudocode, which takes as input an array A and an index i in A). What is the worst-case time complexity of `MaxHeapify` (One or more options are correct)

Algorithm MaxHeapify(A, i):

$l := \text{left}(i)$

$r := \text{right}(i)$

if $l \leq A.\text{heap-size}$ **and** $A[l] > A[i]$ **then**

$largest := l$

else

$largest := i$

if $r \leq A.\text{heap-size}$ **and** $A[r] > A[largest]$ **then**

$largest := r$

if $largest \neq i$ **then**

swap($A[i], A[largest]$)

MaxHeapify($A, largest$)

- ☒ $O(h)$ with h the height of the max-heap
- ☐ $O(n)$ with n the number of keys of the max-heap
- ☐ $O(\log(h))$ with h the height of the max-heap
- ☒ $O(\log(n))$ with n the number of keys of the max-heap
- ☐ $O(n/2)$ with n the number of keys of the max-heap

Q5

We consider two statements.

Statement A is: $2^{(n+1)}$ is in $O(2^n)$.

Statement B is: 2^{2n} is in $O(2^n)$.

Which of the following holds?

- ☐ B is true but A is false.
- ☐ A and B are both true.
- ☒ A is true but B is false.
- ☐ A and B are both false.

Q6

Which of the following recurrence equations(one or more) describe(s) correctly a function T for the worst-case time complexity of merge sort (assuming that $T(1)=1$)?

- ☒ $2T(n/2) + 4n$
- ☐ $T(n/2) + n$
- ☐ $4T(n/2) + 2n$
- ☒ $2T(n/2) + n$
- ☒ $2T(n/2) + cn$ with c a constant

Q7

Why is quicksort a good sorting algorithm?

- ☒ Because it is rather efficient, despite the fact that it does not have an optimal worst case time complexity.
- ☐ Because it is the default sorting algorithm in the libraries of major programming languages.
- ☐ Because it is a recursive algorithm.
- ☐ Because it has good worst-case time complexity.

Q8

Which one of the following arrays is(are) a max-heap?

- ☒ $[7,6,3,5,4,2,1]$
 - ☒ $[7,4,6,3,1,5,2]$
 - ☐ $[7,6,4,1,2,5,3]$
 - ☐ $[7,6,2,5,4,3,1]$
 - ☐ $[1,2,3,4,5,6,7]$
-

Q9

We consider the algorithm partition used in quicksort (see the attached pseudo-code). What is the worst-case time complexity?

```
Algorithm partition( $A, p, r$ ):  
   $x := A[r]$   
   $i := p - 1$   
  for  $j = p$  to  $r - 1$  do  
    if  $A[j] \leq x$  then  
       $i := i + 1$   
      exchange  $A[i]$  with  $A[j]$   
  exchange  $A[i + 1]$  with  $A[r]$   
  return  $i + 1$ 
```

- ☐ The worst-case time complexity is in Theta (n^2) with n the number of elements in the array A .
 - ☐ The worst-case time complexity is in Theta ($n \log(n)$) with n the number of elements in the array A .
 - ☒ The worst-case time complexity is in Theta (n) with n the number of elements in the array A .
 - ☐ The worst-case time complexity is in Theta ($\log(n)$) with n the number of elements in the array A .
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Q10

We consider the algorithm MaxHeapify(see attached pseudo-code) and we consider an execution of MaxHeapify with input an array A of length n , and an index i which is at least $n/2$.

What is the time complexity of such an execution?

Algorithm MaxHeapify(A, i):

$l := \text{left}(i)$

$r := \text{right}(i)$

if $l \leq A.\text{heap-size}$ **and** $A[l] > A[i]$ **then**

$largest := l$

else

$largest := i$

if $r \leq A.\text{heap-size}$ **and** $A[r] > A[largest]$ **then**

$largest := r$

if $largest \neq i$ **then**

swap($A[i], A[largest]$)

MaxHeapify($A, largest$)

- ☐ Approximately $\log(n)$ time, with n the length of the array A .
- ☒ Elementary time.
- ☐ Approximately n time, with n the length of the array A .
- ☐ Approximately $n/2$ time with n the length of the array A .

Q11

Merge-sort is a divide-and-conquer algorithm.

Suppose we adapt merge sort by splitting the sequence into 4 more or less equal-size parts. Do we get an essentially better worst-case time complexity?

- ☒ No, because the height of the recursion tree is then still proportional to $\log(n)$.
 - ☐ Yes, we get an algorithm which is twice as fast because the recursion tree has smaller height.
 - ☐ No, because then the merge-procedure takes more time,
 - ☐ Yes, because the merge-procedure takes smaller inputs.
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Q12

We consider bucket sort (see the attached pseudo-code description).

What is the worst-case running time of bucket sort?

Algorithm bucketSort(A):

$n := A.length$

new array $B[0 \dots n - 1]$

for $i := 0$ **to** $n - 1$ **do**

 make $B[i]$ an empty list

for $i := 1$ **to** n **do**

 insert $A[i]$ into list $B[\lfloor n \cdot A[i] \rfloor]$

for $i := 0$ **to** $n - 1$ **do**

 insertionSort($B[i]$)

concatenate $B[0], B[1], \dots, B[n - 1]$

- ☐ It is in Theta (n^2), because worst-case sorting is always in n^2 .
 - ☒ It is in Theta (n^2), because we may call insertion sort, which is quadratic, on an input of size n , namely if $\lfloor nA[i] \rfloor$ is the same for all indices i .
 - ☐ It is in Theta (n^3), because we may call insertion sort which is quadratic n times.
 - ☐ It is in Theta (n), because the for-loops are not nested.
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Q13

An algorithm with worst-case time complexity Theta($n \log(n)$) is always, for any input, faster than an algorithm with worst-case time complexity in Theta(n^2).

Is this statement true or false, and why?

- ☒ False, because for a small input or for a special input an algorithm with worst-case time complexity in Theta (n^2) may perform better than an algorithm with worst-case time complexity in Theta ($n \log(n)$).
 - ☐ True, because Theta gives a strict bound.
 - ☐ True, because $n \log(n)$ has a lower rate of growth than n^2 .
 - ☐ False, because $n \log(n)$ and n^2 have the same rate of growth anyway.
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Q14

We consider the algorithm for partition (see pseudo-code attached).

Apply partition to the input array $A = [3, 7, 6, 1, 8, 5, 2, 4]$ and indices $p = 1$ and $r = 8$.

What is A after having executed the iterations for $j = 1, 2, 3, 4$?

Algorithm partition(A, p, r):

$x := A[r]$

$i := p - 1$

for $j = p$ **to** $r - 1$ **do**

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

$i := i + 1$

 exchange $A[i]$ with $A[j]$

exchange $A[i + 1]$ with $A[r]$

return $i + 1$

☒ $A = [3, 1, 6, 7, 8, 5, 2, 4]$

☐ $A = [3, 1, 7, 6, 8, 5, 2, 4]$

☐ $A = [3, 7, 6, 4, 8, 5, 2, 1]$

☐ $A = [1, 3, 6, 7, 8, 5, 2, 4]$

Q15

We consider counting sort(see the attached pseudo-code). What happens if we change the last for-loop by letting the index j go from 1 up to $A.length$?

Algorithm countingSort(A, B, k):

new array $C[0 \dots k]$

for $i := 0$ **to** k **do**

$C[i] := 0$

for $j := 1$ **to** $A.length$ **do**

$C[A[j]] := C[A[j]] + 1$

for $i := 1$ **to** k **do**

$C[i] := C[i] + C[i - 1]$

for $j := A.length$ **downto** 1 **do**

$B[C[A[j]]] := A[j]$

$C[A[j]] := C[A[j]] - 1$

- ☐ The change has no effect at all.
 - ☐ The algorithm is no longer correct.
 - ☐ The algorithm sorts in reverse order.
 - ☒ The algorithm still sorts correctly, but is no longer stable.
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Q16

What is the best-case time complexity of insertion sort?

- ☐ Theta (n^2)
 - ☐ Theta ($\log(n)$)
 - ☐ Theta ($n \log(n)$)
 - ☒ Theta (n)
-

Q17

The heapsort algorithm consists of two parts: first build a max-heap and then iterate removing elements from it. Which of the two parts is responsible for the worst case complexity of heapsort?

- ☒ The second part, because the first part is linear.
 - ☐ It depends on the input.
 - ☐ Both parts have the worst time complexity of heapsort.
 - ☐ The first part, because the second part is linear.
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Q18

We consider an application of buildMaxHeap (see attached) to $A = [1, 2, 3, 4, 5, 6, 7]$. What is A after the first iteration of the for-loop?

Algorithm buildMaxHeap(H):

$H.heap-size := H.length$

for $i = \lfloor H.length/2 \rfloor$ **downto** 1 **do**

 MaxHeapify(H, i)

☒ [1,2,7,4,5,6,3]

☐ [1,5,3,4,2,6,7]

☐ [3,2,7,4,5,6,1]

☐ [3,2,1,4,5,6,7]

Q19

Consider the following recurrence equation:

$T(0) = 1$ and $T(n) = (n-1) + n$ for $n > 1$.

What is the big-O of T ?

☐ $T(n)$ in $O(n)$

☒ $T(n)$ in $O(n^2)$

☐ $T(n)$ in $O(2^n)$

☐ $T(n)$ in $O(n \log(n))$

Q20

We consider a max-heap containing n different keys. Why do we have that the height of such a max-heap is approximately $\log(n)$?

- ☐ Because a max-heap is a binary tree.
 - ☐ Because the keys in a max-heap are partially ordered.
 - ☒ Because a max-heap is an almost-complete binary tree.
 - ☐ Because all elements are different.
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Q21

Why does the linear time complexity of counting sort not contradict the lower bound on the worst-case complexity of comparison-based sorting?

- ☐ Because the lower bound is on worst-case and the linear time complexity of counting sort is best-case.
 - ☐ Because the lower bound is linear time.
 - ☒ Because counting sort is not comparison-based.
 - ☐ Because counting sort uses additional memory.
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Q22

Which of the following sorting algorithm has/have a worst-case time complexity in $\Theta(n \log(n))$?

- ☐ counting sort
 - ☒ heapsort
 - ☐ insertion sort
 - ☐ bucket sort
 - ☒ merge sort
 - ☐ selection sort
 - ☐ quicksort
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Q23

We consider the algorithm for the bottom-up max-heap construction (see pseudo-code attached).

What happens if we let the loop-index increase from 1 upwards to floor of $\lfloor A.length/2 \rfloor$?

Algorithm buildMaxHeap(H):

$H.heap-size := H.length$

for $i = \lfloor H.length/2 \rfloor$ **downto** 1 **do**

 MaxHeapify(H, i)

- ☐ Then we need to take the ceiling of $A.length / 2$ instead of the floor in order for the algorithm to be correct.
 - ☐ It works equally well,
 - ☒ Then the algorithm is no longer correct.
 - ☐ Then we should do the recursive call of MaxHeapify on A and $i-1$ instead of on A and i in order for the algorithm to be correct.
-

Q24

Which recurrence equation correctly describes, next to $T(0) = 1$, the worst-case time complexity of quicksort ?

- ☐ $T(n) = 2T(n/2) + cn$
 - ☐ $T(n) = T(n) + T(0) + cn$
 - ☒ $T(n) = T(n-1) + T(0) + cn$
 - ☐ $T(n) = 2T(n/2)$
-

Q25

What is the worst-case time complexity of insertion sort?

- ☐ $O(\log n)$
 - ☒ $\Theta(n^2)$
 - ☐ $\Theta(n \log(n))$
 - ☐ $O(n \log(n))$
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Q26

What is the time complexity of quicksort on an array in which all keys are identical?

- ☐ $\Theta(n \log(n))$, because the array is already sorted, but you still have the recursive calls.
 - ☒ $\Theta(n^2)$, because when partitioning one side will always be empty.
 - ☐ $\Theta(1)$, because the array is trivial.
 - ☐ $\Theta(n)$, because the array is already sorted.
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Q27

We consider insertion sort(see attached). What happens if we swap the order of the tests for the while-loop in line 5?

- ☐ The algorithm then sorts in reverse order.
 - ☒ We may then possibly compare $A[0]$ with key, but $A[0]$ does not exist.
 - ☐ The program then does not terminate.
 - ☐ It does not matter.
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Q28

When is a sorting algorithm said to be stable?

- ☐ If the algorithm is linear time on an array that is already sorted.
 - ☐ If sorting the array twice in a row does not affect the result.
 - ☒ If it respects the order of equal keys in the array.
 - ☐ If it does not need extra memory next to the input array.
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Q29

What is the running time of heapsort if the input is an array that is already sorted (in increasing order)?

- ☐ The running time is then in $O(\log(n))$.
- ☐ The running time is then in $O(n)$.
- ☐ The running time is then in $O(n^2)$.
- ☒ The running time is then in $O(n \log(n))$.