

MONASH INFORMATION TECHNOLOGY

FIT2004 Algorithms and Data Structures

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Referencing materials by Nathan Companez, Aamir Cheema, Arun Konagurthu and Lloyd Allison







Ready?

Agenda

- Complexity Analysis
 - Big-O
 - Recurrence relation



Agenda

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 - Big-O
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Agenda

Complexity Analysis

- Big-O

Recurrence relation

Covered in Tutorial 02 using tutorial questions as case study





Let us begin...

Recap



You have done time complexity last time



- You have done time complexity last time
- You know what Big-O is

Recap



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What is the complexity of an algorithm in Big-O notation that runs in 30N $log(N^2) + 10 log N + 8N$?

- A. O(N log N)
- B. O(N log (N²))
- c. $O(N \log (N^2) + N + \log N)$
- D. Option D because





Yes I finally got the answer it's 637,159.017





looks at choices

A) 12 B) 21

C) 21.5

D) 12.5



What is your answer?

Recap



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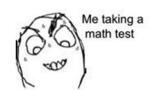
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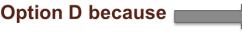
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- **Option D because**





Yes I finally got the answer it's 637,159,017





looks at choices

D) 12.5



 $Log N^2 = 2 log N$



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- If I have a list, and I want to sort it with bubble sort...



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- If I have a list, and I want to sort it with bubble sort...
 - Best case?
 - Worst case?



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- If I have a list, and I want to sort it with bubble sort...
 - Best case? O(1) when list is empty
 - Worst case? O(N^2) when list is sorted in reverse order



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Recap



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- You know what Big-O is

n > k, n is larger and significantly larger than k

big-O notation not care small output

- If I have a list, and I want to sort it with bubble sort...
 - Best case? O(N) when list is sorted and we can terminate earlier
 - Worst case? O(N^2) when list is sorted in reverse order

N > k Big-O complexity only cares about the large N scalable part



Recap



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you vs. the guy she tells you not to worry about



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- You know what Big-O is
- So what's new?



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 - Time complexity for recursion



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 - By solving recurrence relation



- You have done time complexity last time
- You know what Big-O is
- So what's new?
 - Space complexity
 - We'll leave this for Lecture 02 when we go through some sorting algorithms
 - Time complexity for recursion
 - By solving recurrence relation



Questions?



- This is asked a lot
 - Final exam



- This is asked a lot
 - Final exam
- Helps you figure out the complexity of recursive functions

Recurrence relation



Look at the algorithm written in Python



- Look at the algorithm written in Python
- Note: Can you explain what this function do? ... and reason out the complexity?
 - Using FIT1008 knowledge



- Look at the algorithm written in Python
- What is the recurrence relation?



- Look at the algorithm written in Python
- What is the recurrence relation? T(N)
 - Base case
 - Recursive case



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- What is the recurrence relation? T(N)
 - Base case T(0) = a
 - Recursive case

Recurrence relation



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```
T(0) = a
T(1) = b
```

Recursive case

Recurrence relation



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```
T(0) = a
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Constant operating cost...
```

Recursive case

Recurrence relation



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Constant operating cost...
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– Recursive case T(N) = T(N-1)

Recurrence relation



- Look at the algorithm written in Python
- What is the recurrence relation? T(N)
 - Base case

```
T(0) = a
T(1) = b
Constant operating cost...
```

- Recursive case
T(N) = T(N-1) * X?

Recurrence relation



- Look at the algorithm written in Python
- What is the recurrence relation? T(N)
 - Base case

```
T(0) = a
T(1) = b
Constant operating cost...
```

Recursive case

```
T(N) = T(N-1) + c constant constant operating cost in multiplying from x * power (x, n-1) Cause of constant operating cost in multiplying...
```

just a simple constant constant operating cost in multiplying

Recurrence relation



- Look at the algorithm written in Python
- What is the recurrence relation? T(N)
 - Base case

```
T(0) = a
T(1) = b
Constant operating cost...
```

Recursive case (general case)
 T(N) = T(N-1) + c
 Cause of constant operating cost in multiplying...



- T(0) = a
- T(1) = b
- T(N) = T(N-1) + c

Recurrence relation



- T(0) = a
- T(1) = b
- T(N) = T(N-1) + c

Now solve it for the complexity



- T(0) = a
- T(1) = b
- T(N) = T(N-1) + c
- Now solve it for the complexity
 - T(N) = T(N-1) + c
 - so T(N-1) = T(N-2) + c



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 - T(N) = T(N-2) + c + c
 - T(N) = T(N-3) + c + c + c

Recurrence relation



- T(0) = a
- T(1) = b
- T(N) = T(N-1) + c

- Now solve it for the complexity

```
- T(N) = T(N-1) + c
```

$$-$$
 so $T(N-1) = T(N-2) + c$

$$- T(N) = T(N-2) + c + c$$

$$- T(N) = T(N-3) + c + c + c$$

Known as telescoping



- T(0) = a
- T(1) = b
- T(N) = T(N-1) + c
- Now solve it for the complexity

$$- T(N) = T(N-1) + c$$

$$-$$
 so T(N-1) = T(N-2) + c

$$- T(N) = T(N-2) + c + c$$

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- T(0) = a
- T(1) = b
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- Now solve it for the complexity
 - T(N) = T(N-1) + c
 - so T(N-1) = T(N-2) + c
 - T(N) = T(N-2) + c + c
 - T(N) = T(N-3) + c + c + c
 - generalized into T(N) = T(N-k) + kc
 - Base case when N=0 or N=1



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 - N-k = 0, therefore base case when k=N. we replace into above...



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 - T(N) = T(N-1) + c
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 - generalized into T(N) = T(N-k) + kc
 - Base case when N=0 or N=1
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 - T(N) = T(N-N) + Nc = T(0) + Nc = a + Nc



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- T(1) = b
- T(N) = T(N-1) + c
- Now solve it for the complexity
 - T(N) = T(N-1) + c
 - so T(N-1) = T(N-2) + c
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 - T(N) = T(N-3) + c + c + c
 - generalized into T(N) = T(N-k) + kc
 - Base case when N=0 or N=1
 - N-k = 0, therefore base case when k=N. we replace into above...
 - T(N) = T(N-N) + Nc = T(0) + Nc = a + Nc = O(N), eliminating the constant



Questions?



- Let us try another one
- Come up with the recurrence relation



- Let us try another one
- Come up with the recurrence relation
- Fun fact, this was a programming question for the exam...



- Let us try another one
- Come up with the recurrence relation
- Fun fact, this was a programming question for the exam...
- Do you understand the code?

Recurrence relation



- Let us try another one
- Come up with the recurrence relation
- Fun fact, this was a programming question for the exam...
- Do you understand the code?

```
power_squaring(x,n):
Returns x^n
via exponential by squaring
if n == 0:
    return 1
elif n == 1:
    return x
# N^4 = N^2 * N^2
elif n%2 == 0:
    return power (x*x, n//2)
# N^9 = N^4 & N^4 * N
elif n%2 == 1:
    return power (x*x, n//2) * n
```

Should be *x



•
$$T(0) = a$$



- T(0) = a
- T(1) = b



- T(0) = a
- T(1) = b
- T(N) = T(N//2) + c when N even



- T(0) = a
- T(1) = b
- T(N) = T(N//2) + c when N even
- T(N) = T(N//2) + d when N odd

Recurrence relation



- T(0) = a
- T(1) = b
- T(N) = T(N//2) + c when N even
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this is output we just care about complexity

— Not that it isn't T(N) = T(N//2) * N WHY?



- T(0) = a
- T(1) = b
- T(N) = T(N//2) + c when N even
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- Now you solve it...



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- Now you solve it...
 Let me take the even one to start



- T(0) = a
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```
- T(N) = T(N//2) + c
```

$$- T(N) = T(N//4) + 2c$$

$$- T(N) = T(N//8) + 3c$$

Recurrence relation



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- T(N) = T(N//4) + 2c
- T(N) = T(N//8) + 3c can you see the pattern?
- $T(N) = T(N//2^k) + kc$



- T(0) = a
- T(1) = b
- T(N) = T(N//2) + c when N even
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- Now you solve it...Let me take the even one to start

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- $T(N) = T(N//2^k) + kc$
- Base when $N//2^k = 1...$ thus $N = 2^k$



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- T(N) = T(N//2) + c when N even
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- $T(N) = T(N//2^k) + kc$
- Base when $N//2^k = 1...$ thus $N = 2^k$ which is $k = \log N$



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- T(N) = T(N//8) + 3c can you see the pattern?
- $T(N) = T(N//2^k) + kc = T(1) + log N c = b + log N c$
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- T(N) = T(N//4) + 2c
- T(N) = T(N//8) + 3c can you see the pattern?
- $T(N) = T(N//2^k) + kc = T(1) + log N c = b + log N c = O(log N)$
- Base when $N//2^k = 1$... thus $N = 2^k$ which is $k = \log N$



Questions?



- Both functions we saw just now are similar
- They are power functions x^n



- Both functions we saw just now are similar
- They are power functions x^n
- But we also saw how their complexity differs
 - O(N)
 - O(log N)



- Both functions we saw just now are similar
- They are power functions x^n
- But we also saw how their complexity differs
 - O(N) = normal power
 - O(log N) = exponential by squaring



Questions?

Recurrence relation



So now you know why functions have such complexity?





- So now you know why functions have such complexity?
- Some of the other common ones...

Recurrence relation:

$$T(N) = T(N/2) + c$$

 $T(1) = b$

Example algorithm?

Binary search

Solution:

O(log N)

Recurrence relation:

$$T(N) = T(N-1) + c$$

 $T(1) = b$

Example algorithm?

Linear search

Solution:

O(N)

Recurrence relation:

$$T(N) = 2*T(N/2) + c*N$$

 $T(1) = b$

Example algorithm?

Merge sort

Solution:

 $O(N \log N)$

Recurrence relation:

$$T(N) = T(N-1) + c*N$$

 $T(1) = b$

Example algorithm?

Selection sort

Solution:

 $O(N^2)$

Recurrence relation:

$$T(N) = 2*T(N-1) + c$$

 $T(0) = b$

Example algorithm?

Naïve recursive Fibonacci

Solution:

$$O(2^N)$$



- So now you know why functions have such complexity?
- Some of the other common ones...
- Exam would also ask you to proof by induction for complexity...



- So now you know why functions have such complexity?
- Some of the other common ones...
- Exam would also ask you to proof by induction for complexity...
 - We will discuss more in the tutorial
 - Now, let us try to make my life difficult in the zoom session later this week...



Questions?



Thank You