

COMP 250

Lecture 33

recurrences 1

March. 30, 2022

What's left to do ?

- Lecture 33, 34 : Recurrences
- Lecture 35, 36, 37: Asymptotic Complexity

Let $t(n)$ be the time or number of instructions to execute an algorithm.

We have discussed how to determine $t(n)$ when our algorithms only have loops.

Let's briefly review a few examples...

Grade School Addition

```
carry = 0  $c_1$   
for  $i = 0$  to  $N - 1$  do  
     $sum \leftarrow a[i] + b[i] + carry$   
     $r[i] \leftarrow sum \% 10$   
     $carry \leftarrow sum / 10$   $c_2N$   
end for  
 $r[N] \leftarrow carry$   $c_3$ 
```

The number of steps is $t(N) = (c_1 + c_3) + c_2 * N$.

When we analyze algorithms with $O()$, we ignore the constants.

Grade School Multiplication

```
Step 1 { for  $j = 0$  to  $N - 1$  do  
         $carry \leftarrow 0$   $N$   
        for  $i = 0$  to  $N - 1$  do  
             $prod \leftarrow (a[i] * b[j] + carry)$   $N^2$   
             $table[j][i + j] \leftarrow prod \% 10$   
             $carry \leftarrow prod / 10$   
        end for  
         $table[j][N + j] \leftarrow carry$   $N$   
    end for  
Step 2 {  $carry \leftarrow 0$   $1$   
        for  $i = 0$  to  $2 * N - 1$  do  
             $sum \leftarrow carry$   $N$   
            for  $j = 0$  to  $N - 1$  do  
                 $sum \leftarrow sum + table[j][i]$   $N^2$   
            end for  
             $r[i] \leftarrow sum \% 10$   $N$   
             $carry \leftarrow sum / 10$   
        end for
```

Similarly, we could include constant factors for each of these terms.

The number of steps is $t(N) = c_0 + c_1 N + c_2 * N^2$.

Selection Sort

```
for  $i = 0$  to  $N - 2$ {
```

```
    index =  $i$ 
    minValue = list[ $i$ ] }  $\sim N$ 
```

```
    for  $k = i + 1$  to  $N - 1$  { // nested loop
```

```
        if ( list[ $k$ ] < minValue ){
            minValue = list[ $k$ ]
            index =  $k$ 
        }
    }
```

```
    if ( index  $\neq i$  )
        list.swap( $i$ , index) }  $\sim N$ 
}
```

The number of steps is $t(N) = c_0 + c_1 N + c_2 * N^2$.

Let $t(n)$ be the time or number of instructions to execute an algorithm.

We have discussed how to determine $t(n)$ when our algorithms only have loops.

Q: How do we determine $t(n)$ for **recursive** algorithms ?

A: We use *recurrence relations*.

Recurrence Relation

A recurrence relation is an equation that defines a sequence of numbers whose n -th term depends on previous terms.

e.g. Fibonacci $F(n) = F(n - 1) + F(n - 2)$

We will consider recurrence relations for time complexity $t(n)$
e.g. the number of steps to execute a recursive algorithm as a function of the *input size* n .

The recurrence expresses $t(n)$ in terms of a smaller input size.

Example 1 : Suppose a list has n elements.

What is $t(n)$ for reversing the list as follows ?

```
reverse( list ){  
    if list.size > 1 {  
        firstElement = list.removeFirst( )  
        reverse( list )  
        list.addLast( firstElement )  
    }  
}
```

Example 1 : Suppose a list has n elements.

What is $t(n)$ for reversing the list as follows ?

```
reverse( list ){  
    if list.size > 1 {  
        firstElement = list.removeFirst()  
        reverse( list )  
        list.addLast( firstElement )  
    }  
}
```

Constant time to do the base case check.

If a linked list is used, then it is constant time for both `removeFirst()` and `addLast()`. But if an arraylist were used, then `removeFirst()` would not be constant time.

$$t(n) = c + t(n - 1)$$

Solving a recurrence using “back substitution”

$$t(n) = c + t(n - 1)$$

Solving a recurrence using back substitution

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

Solving a recurrence using back substitution

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

Solving a recurrence using back substitution

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

↙
e.g. base case is $n = 1$
when reversing a list

Example 2 : Sorting a list

```
sort( list ) {  
    if list.size == 1  
        return list  
    else{  
        minElement = list.removeMin()  
        sort(list)  
        return list.addFirst( minElement )  
    }  
}
```

What is the recurrence relation?

Example 2 : Sorting a list

```
sort( list ) {  
  if list.size == 1 ← Base case  $n = 1$ .  
    return list      (But we do this test everytime.)  
  else{  
    minElement = list.removeMin() ← Time proportional to  $n$   
    sort(list)  
    return list.addFirst( minElement )  
  }  
}
```

Depends on list data structure (worst case is proportional to n)

$$t(n) = c_1 + c_2 n + t(n - 1)$$

Let's solve the slightly simpler recurrence, by dropping the constant term.

$$t(n) = c n + t(n - 1)$$



Let's solve the slightly simpler recurrence.

$$t(n) = c n + t(n - 1)$$

$$= c n + c \cdot (n - 1) + t(n - 2)$$



Let's solve the slightly simpler recurrence. (We are sorting a list. Base case is $n = 1$.)

$$t(n) = c n + t(n - 1) \quad k = 0$$

$$= c n + c \cdot (n - 1) + t(n - 2) \quad k = 1$$

$$= \dots$$

$$= c \{ n + (n - 1) + (n - 2) + \dots + (n - k) \} + t(n - k - 1)$$



Let's go all the way to the end, and use a cleaner base case $t(0)$ or $n - k = 1$.


Let's solve the slightly simpler recurrence, and cleaner base case ($n = 0$).

$$t(n) = c n + t(n - 1)$$

$$= c n + c \cdot (n - 1) + t(n - 2)$$

$$= \dots$$

$$= c \{ n + (n - 1) + (n - 2) + \dots + (n - k) \} + t(n - k - 1)$$

$$= c \{ n + (n - 1) + (n - 2) + \dots + 2 + 1 \} + t(0)$$


Let's solve the slightly simpler recurrence, **and cleaner base case** ($n = 0$).

$$t(n) = cn + t(n - 1)$$

$$= cn + c \cdot (n - 1) + t(n - 2)$$

$$= \dots$$

$$= c \{ n + (n - 1) + (n - 2) + \dots + (n - k) \} + t(n - k - 1)$$

$$= c \{ n + (n - 1) + (n - 2) + \dots + 2 + 1 \} + t(0)$$

$$= \frac{cn(n + 1)}{2} + t(0)$$

Example 3: Tower of Hanoi

```
tower(n, start, finish, other){    // base case is n=1, so t(1) = 1
  if n == 1
    move from start to finish
  else {
    tower( n-1, start, other, finish)
    move from start to finish
    tower( n-1, other, finish, start)
  }
}
```


Q: How many moves are needed for a tower with n disks ?

$$t(n) = 1 + 2 t(n - 1)$$


What do you think the solution will be ?

$$t(n) \sim n^2 ? \quad n^3 ? \quad 2^n ?$$

Tower of Hanoi recurrence

$$t(n) = 1 + 2 t(n - 1)$$


Tower of Hanoi recurrence

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


Tower of Hanoi recurrence

$$\begin{aligned}t(n) &= 1 + 2 t(n - 1) \\ &= 1 + 2(1 + 2 t(n - 2)) \\ &= (1 + 2) + 4 t(n - 2)\end{aligned}$$

Tower of Hanoi recurrence

$$\begin{aligned}t(n) &= 1 + 2 t(n - 1) \\ &= 1 + 2(1 + 2 t(n - 2)) \\ &= (1 + 2) + 4 t(n - 2) \\ &= (1 + 2) + 4 (1 + 2 t(n - 3))\end{aligned}$$

Tower of Hanoi recurrence

$$\begin{aligned}t(n) &= 1 + 2 t(n - 1) \\ &= 1 + 2(1 + 2 t(n - 2)) \\ &= (1 + 2) + 4 t(n - 2) \\ &= (1 + 2) + 4 (1 + 2 t(n - 3)) \\ &= (1 + 2 + 4) + 8 t(n - 3)\end{aligned}$$

Tower of Hanoi recurrence

$$\begin{aligned}t(n) &= 1 + 2 t(n - 1) && k = 1 \\&= 1 + 2(1 + 2 t(n - 2)) \\&= (1 + 2) + 4 t(n - 2) && k = 2 \\&= (1 + 2) + 4 (1 + 2 t(n - 3)) \\&= (1 + 2 + 4) + 8 t(n - 3) && k = 3 \\&= \dots \\&= (1 + 2 + 4 + 8 + \dots + 2^{k-1}) + 2^k t(n - k)\end{aligned}$$

Tower of Hanoi recurrence

$$\begin{aligned}t(n) &= 1 + 2 t(n - 1) \\&= 1 + 2(1 + 2 t(n - 2)) \\&= (1 + 2) + 4 t(n - 2) \\&= (1 + 2) + 4 (1 + 2 t(n - 3)) \\&= (1 + 2 + 4) + 8 t(n - 3) \\&= \dots \\&= (1 + 2 + 4 + 8 + \dots + 2^{k-1}) + 2^k t(n - k) \\&= (1 + 2 + 4 + 8 + \dots + 2^{n-2}) + 2^{n-1} t(1)\end{aligned}$$

$k = n - 1$

verify

Tower of Hanoi recurrence

$$\begin{aligned}t(n) &= 1 + 2 t(n - 1) \\&= 1 + 2(1 + 2 t(n - 2)) \\&= (1 + 2) + 4 t(n - 2) \\&= (1 + 2) + 4 (1 + 2 t(n - 3)) \\&= (1 + 2 + 4) + 8 t(n - 3) \\&= \dots \\&= (1 + 2 + 4 + 8 + \dots + 2^{k-1}) + 2^k t(n - k) \\&= (1 + 2 + 4 + 8 + \dots + 2^{n-2}) + 2^{n-1} t(1) \\&= (2^{n-1} - 1) + 2^{n-1} t(1)\end{aligned}$$

Tower of Hanoi recurrence

$$\begin{aligned}t(n) &= 1 + 2 t(n - 1) \\&= 1 + 2(1 + 2 t(n - 2)) \\&= (1 + 2) + 4 t(n - 2) \\&= (1 + 2) + 4 (1 + 2 t(n - 3)) \\&= (1 + 2 + 4) + 8 t(n - 3) \\&= \dots \\&= (1 + 2 + 4 + 8 + \dots + 2^{k-1}) + 2^k t(n - k) \\&= (1 + 2 + 4 + 8 + \dots + 2^{n-2}) + 2^{n-1} t(1) \\&= (2^{n-1} - 1) + 2^{n-1} t(1) \\&= 2^n - 1, \text{ since } t(1) = 1 \longleftarrow \text{Base case for Tower of Hanoi}\end{aligned}$$

You should know by now....

$$1 + 2 + 3 + \dots + k = ?$$

$$1 + 2 + 4 + 8 + \dots + 2^k = ?$$

$$1 + x + x^2 + x^3 + \dots + x^k = ?$$

Example 4: Binary Search

```
binarySearch( list, value, low, high ){           //  n = high - low + 1
  if low <= high {
    mid = (low + high) / 2
    if value == list[mid]
      return value
    else if value < list[mid]
      return binarySearch(list, value, low,  mid - 1 )
    else
      return binarySearch(list, value, mid+1, high)
  }
  else
    return -1
}
```

$$t(n) = c + t\left(\frac{n}{2}\right)$$

The list sizes might not be exactly $n/2$, but will be at most off by 1.

Example 4: Binary Search

```
binarySearch( list, value, low, high ){      //  n = high - low + 1
  if low <= high {
    mid = (low + high) / 2
    if value == list[mid]
      return value      ← Base case : found value
    else if value < list[mid]
      return binarySearch(list, value, low,  mid - 1 )
    else
      return binarySearch(list, value, mid+1, high)
  }
  else
    return -1      ← Base case : high < low,    t(1) = 1
}
```

$$t(n) = c + t\left(\frac{n}{2}\right), \quad n > 1$$

Binary search recurrence

$$t(n) = c + t\left(\frac{n}{2}\right)$$

Note this doesn't quite capture the situation, since mid is excluded. But close enough!

Binary search recurrence

$$\begin{aligned}t(n) &= c + t\left(\frac{n}{2}\right) \\ &= c + c + t\left(\frac{n}{4}\right)\end{aligned}$$

Binary search recurrence

$$\begin{aligned}t(n) &= c + t\left(\frac{n}{2}\right) \\ &= c + c + t\left(\frac{n}{4}\right) \\ &= c + c + \cdots + t\left(\frac{n}{2^k}\right)\end{aligned}$$

Binary search recurrence

$$t(n) = c + t\left(\frac{n}{2}\right)$$

$$= c + c + t\left(\frac{n}{4}\right)$$

$$= c + c + \dots + t\left(\frac{n}{2^k}\right)$$

$$= c + c + \dots + c + t\left(\frac{n}{n}\right)$$

For the purpose of solving the recurrence, we assume that n is a power of 2.



$$n = 2^k$$

Binary search recurrence

$$t(n) = c + t\left(\frac{n}{2}\right)$$

$$= c + c + t\left(\frac{n}{4}\right)$$

$$= c + c + \dots + t\left(\frac{n}{2^k}\right)$$

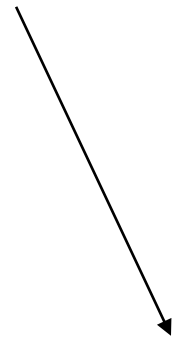
$$= c + c + \dots + c + t\left(\frac{n}{n}\right)$$

$$= c \log_2 n + t(1)$$



Base case

For the purpose of solving the recurrence, we assume that n is a power of 2.



$$n = 2^k$$

Today's Recurrences

$$t(n) = c + t(n - 1)$$

$$t(n) = cn + t(n - 1)$$

$$t(n) = c + 2t(n - 1)$$

$$t(n) = c + t\left(\frac{n}{2}\right)$$

Coming up...

Lectures

Fri, April 1

more recurrences

Mon, Wed, Fri : April 4, 6, 8

big O, big Theta, big Omega

best and worst cases

Assessments

Quiz 5 is in Mon. April 4

Practice quizzes are available for Quiz 5 and for lectures after that (33-37)

Assignment 4 due Wed. April 6.