Lecture 10

Mathematical Analysis of Recursive Algorithms and Solving Recurrences: Iteration Method, Substitution Method





MergeSort Algorithm Process

Here is one way to implement merge:

- Create an empty list called the <u>result list</u>.
- Do the following until one of the input lists is empty:
 - Remove the first element of the list that has a lesser first element and append it to the result list.
- When one of the lists is empty, append all elements of the other list to the result.

Use the merge algorithm to find the third step of the merge of A and B. Here are the first two steps:

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Initial State	A = [2,4,9]	B = [1,7,13,15]	Results = []
First Step	A = [2,4,9]	B = [7,13,15]	Results = [1]
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Second Step	A =[4,9]	B = [7,13,15]	Results = [1,2]

A = [9] B = [7,13,15] Results = [1,2]

A = [9] B = [13,15] Results = [1,2,4,7]

A = [] B = [] Results = [1,2,4,7,9,13,15]

A = [9] B = [7,13,15] Results = [1,2,4]



MergeSort Algorithm Process

> For the third step, you compare the smallest of the first (smallest) elements of A and B, move this element over to Results. In this case, the smaller value is 4 (in element A), which gets moved to the end of Results.

$$A = [9], B = [7, 13, 15],$$
Results $= [1, 2, 4]$

The 3rd STEP of MERGE algorithm for merging A and B is as below

Initial State	A = [2,4,9]	B = [1,7,13,15]	Results = []
First Step	A = [2,4,9]	B = [7,13,15]	Results = [1]
Second Step	A =[4,9]	B = [7,13,15]	Results = [1,2]

The correct of the 3rd STEP of merging A and B is



A = [9] B = [7,13,15] Results = [1,2,4]

MergeSort Steps to Implement in Python



```
def merge(left, right):
    result = []
   left_idx, right_idx = 0, 0
   while left_idx < len(left) and right_idx < len(right):
        # change the direction of this comparison to change the direction of the sort
        if left[left_idx] <= right[right_idx]:</pre>
           result.append(left[left_idx])
           left_idx += 1
        else:
            result.append(right[right_idx])
           right_idx += 1
   if left:
        result.extend(left[left_idx:])
   if right:
        result.extend(right[right_idx:])
    return result
def merge_sort(m):
    if len(m) <= 1:
        return m
   middle = len(m) // 2
   left = m[:middle]
   right = m[middle:]
   left = merge_sort(left)
   right = merge_sort(right)
   return list(merge(left, right))
```



Recurrence relation for complexity analysis

- > Already discussed the way to perform analysis of loops
- Many algorithms are recursive, so recurrence relation for time complexity.
- > We find running time on input size n (smaller sizes) as a function of n, e.g., Merge Sort
 - In merge sort, array is divided in two halves with recursive repetition until to get merge results
 - Thus, T(n)=2T(n/2)+cn
 - Other algorithms are Binary search, Tower of Hanoi etc.

Information about Sorting Algorithms:

<u>3-way Merge Sort</u>, <u>Selection Sort</u>, <u>Bubble Sort</u>, <u>Insertion Sort</u>, <u>Merge Sort</u>, <u>Heap Sort</u>, <u>QuickSort</u>, <u>Radix Sort</u>, <u>Counting Sort</u>, <u>Bucket Sort</u>, <u>ShellSort</u>, <u>Comb Sort</u>

Recurrence Methods

- 1. Iteration Method
- 2. Substitution Method
- 3. Recurrence Tree Method
- 4. Master Theorem





The Iteration Method

- > Convert the recurrence into a summation and try to bound it using known series
 - Iterate the recurrence until the initial condition is reached.
 - Use back-substitution to express the recurrence in terms of n and the initial (boundary) condition.



The Iteration Method (Cont !!!)

$$T(n) = c + T(n/2)$$
 $T(n/2) = c + T(n/4)$
 $T(n/2) = c + T(n/4)$
 $T(n/4) = c + T(n/8)$
 $T(n/4) = c + T(n/8)$
 $T(n/4) = c + T(n/8)$

Assume $n = 2^k$
 $T(n) = c + c + ... + c + T(1)$
 $n/2^i = 1, = n = 2^i = n = 2^i = n/2 = n/2 = n = 2^i = n/2 = n/$



Iteration Method – Example

$$T(n) = n + 2T(n/2)$$

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

$$= n + n + 4T(n/4)$$

$$= n + n + 4(n/4 + 2T(n/8))$$

$$= n + n + n + 8T(n/8)$$
... = $in + 2^{i}T(n/2^{i})$

$$= kn + 2^{k}T(1)$$

$$= nlgn + nT(1) = \Theta(nlgn)$$



Substitution Method

- The substitution method for solving recurrences can be described in two steps:
 - Guess the form of the solution.
 - Use induction to show that the guess is valid.
- > This method is especially powerful when we encounter recurrences that are non-trivial and unreadable via the master theorem.
- > Substitution method can be used to establish both upper and lower bounds on recurrences.
- > The name comes from the substitution of the guessed answer for the function when the inductive hypothesis is applied to smaller values.
- > This method is powerful, but it is only applicable to instances where the solutions can be guessed.



Substitution method (Cont !!!)

- > Guess a solution
 - T(n) = O(g(n))
 - Induction goal: apply the definition of the asymptotic notation
 - $T(n) \le d g(n)$, for some d > 0 and $n \ge n_0$
 - *Induction hypothesis:* $T(k) \le d g(k)$ *for all* k < n
- > Prove the induction goal
 - Use the induction hypothesis to find some values of the constants d and n_0 for which the induction goal holds



Substitution Method: Example

- > Consider the recurrence T(n) = 2T(n/2) + n to show that is $O(n \lg(n))$.
- > Solution:
 - To prove $T(n) \le cn \lg(n)$, we assume that bound holds for $\frac{n}{2}$.
 - $-T(n) \le c(\frac{n}{2}) \lg(n/2) + n$
 - $\leq cn lg(n/2) + n$
 - $\leq cn lg(n) cn lg(2) + n$
 - $\leq cn \, lg(n) cn \, lg(2) + n$ $\leq cn \, lg(n) + n(1 clg(2))$ $\leq cn \, lg(n) + n(1 c)$

 - $\leq cn \lg(n)$, for any c>1. we are done with it.

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Recurrence Relation [T(n)= n*T(n-1)] Substitution Method

$$T(n) = \begin{cases} 1 & \text{if } n = 1 \dots Base Condition} \\ n * T(n-1) & \text{if } n > 1 \dots recurrence relation} \end{cases}$$

> Solution

$$-T(n) = n * T(n-1) \longrightarrow (1)$$

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

$$-T(n-1) = (n-1) * T((n-2)) \longrightarrow (2)$$

$$-T(n-2) = (n-2) * T(n-3) \longrightarrow (3)$$

- Substitute (3) in (2) and then (2) in (1) to know the trend as below

$$-T(n) = n * (n-1) * (n-2)T(n-3)$$
 To eliminate $T(n-3)$ take it upto n-1 steps.

$$-T(n) = n * (n-1) * (n-2) * (n-3) ... T(n-(n-1))$$

$$- T(n) = n * (n-1) * (n-2) * (n-3) \dots T(n-n+1)$$

$$-T(n) = n * (n-1) * (n-2) * (n-3) ... T(1)$$

$$- T(n) = n * (n-1) * (n-2) * (n-3) \dots 3 * 2 * 1$$

$$- \qquad = n * n \left(1 - \frac{1}{n}\right) * n \left(1 - \frac{2}{n}\right) * n \left(1 - \frac{3}{n}\right) \dots n \left(\frac{3}{n}\right) * n \left(\frac{2}{n}\right) * n \left(\frac{1}{n}\right)$$

 $-T(n) = O(n^n)$ which is a factorial time multiplication



Example: Binary Search

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

- \rightarrow Guess: T(n) = O(lgn)
 - Induction goal: $T(n) \le d \lg n$, for some d and $n \ge n_0$
 - Induction hypothesis: $T(n/2) \le d \lg(n/2)$
- > Proof of induction goal:

$$T(n) = T(n/2) + c \le d \lg(n/2) + c$$

= $d \lg n - d + c \le d \lg n \text{ if: } -d + c \le 0, d \ge c$

Example

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

- \rightarrow Guess: $T(n) = O(n^2)$
 - Induction goal: $T(n) \le c n^2$, for some c and $n \ge n_0$
 - *Induction hypothesis:* $T(n-1) \le c(n-1)^2$ *for all* k < n
- > *Proof of induction goal:*

$$T(n) = T(n-1) + n \le c (n-1)^2 + n$$

$$= cn^2 - (2cn - c + n) \le cn^2$$

$$if: 2cn - c + n \ge 0 \Leftrightarrow c \ge n/(2n-1) \Leftrightarrow c \ge 1/(2-1/n)$$

- For $n \ge 1 \Longrightarrow 2 - 1/n \ge 1 \Longrightarrow any \ c \ge 1$ will work

Example

$$T(n) = 3T(n/4) + cn^2$$

- \rightarrow Guess: $T(n) = O(n^2)$
 - Induction goal: $T(n) \le dn^2$, for some d and $n \ge n_0$
 - Induction hypothesis: $T(n/4) \le d (n/4)^2$
- > Proof of induction goal:

$$T(n) = 3T(n/4) + cn^{2}$$

$$\leq 3d (n/4)^{2} + cn^{2}$$

$$= (3/16) d n^{2} + cn^{2}$$

$$\leq d n^{2} \quad \text{if: } d \geq (16/3)c$$

> Therefore: $T(n) = O(n^2)$

Substitution Method: Example



> void RecRel (int n)

$$-IFn > 1$$

- > RecRel(n/2)
- > RecRel(n/2)

$$T(n) = 2T\left(\frac{n}{2}\right) + n2T\left(\frac{n}{2}\right) + n \quad --- \to (1)$$

$$= 2\left[2T\left(\frac{n}{2^{2}}\right) + \frac{n}{2}\right] + n$$

$$= 2^{2}\left[2T\left(\frac{n}{2^{2}}\right) + n + n \quad ---- \to (2)\right]$$

$$= 2^{2}\left[2T\left(\frac{n}{2^{3}}\right) + \frac{n}{2^{2}}\right] + 2n$$

$$= 2^{3}\left(T\left(\frac{n}{2^{3}}\right)\right) + 3n \quad ---- \to (3)$$

$$T(n) = 2^{k}\left(T\left(\frac{n}{2^{k}}\right)\right) + kn$$

$$Assume$$

$$T\left(\frac{n}{2^{k}}\right) = T(1)$$

$$\therefore n/2^{k} = 1, n = 2^{k} \to k = \log n$$

$$T(n) = 2^{k}T(1) + kn$$

$$T(n) = nX1 + n \log n = \mathbf{0}(n)$$

$$T(n) = \begin{cases} 1 & \text{if } n = 1 \\ 2T\left(\frac{n}{2}\right) + n & \text{if } n > 1 \end{cases}$$

Assume
$$T\left(\frac{n}{2^{k}}\right) = T(1)$$

$$\therefore n/2^{k} = 1, n=2^{k} \implies k=\log n$$

$$T(n) = 2^{k}T(1) + kn$$

$$T(n) = nX1 + n\log n = \mathbf{O}(n\log n)$$

Thank You!!!

Have a good day

