Algorithms and Problem Solving (15B11CI411) EVEN 2022



Module 1: Lecture 3

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Recurrences and Running Time

 An equation or inequality that describes a function in terms of its value on smaller inputs.

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

- Recurrences arise when an algorithm contains recursive calls to itself
- What is the actual running time of the algorithm?
- Need to solve the recurrence

Example Recurrences

$$T(n) = T(n-1) + n$$
 $O(n^2)$

· Recursive algorithm that loops through the input to eliminate one item

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

Recursive algorithm that halves the input in one step

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

Recursive algorithm that halves the input but must examine every item in the input

$$T(n) = 2T(n/2) + 1$$
 $O(n)$

 Recursive algorithm that splits the input into 2 halves and does a constant amount of other work

Analysis of BINARY-SEARCH

T(n) – running time for an array of size n

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Alg.: BINARY-SEARCH (A, first, last, x)
     if (first > last)
                                                         constant time: c<sub>1</sub>
         return FALSE
     mid \leftarrow \lfloor (first + last)/2 \rfloor
                                                         constant time: c_2
     if x = A[mid]
                                                         constant time: c_3
         return TRUE
     if (x < A[mid])
         BINARY-SEARCH (A, first, mid-1, x)
                                                          same problem of size n/2
     if (x > A[mid])
         BINARY-SEARCH (A, mid+1, last, x)
                                                         same problem of size n/2
 T(n) = c + T(n/2)
```

How to solve Recurrence Relations

- Substitution method
- Iteration method
- · Recursion Tree method
- Master method

Substitution Method

- 1. Guess the form of the solution.
- 2. Use mathematical induction to find the constants and show that the solution works.

Substitution method

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

Example 1:T(n) = T(n-1) + n

- 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:

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

$$\leq c (n-1)^{2} + n$$

$$= cn^{2} - (2cn - c - n)$$

$$\leq cn^{2} if: 2cn - c - n \geq 0 \Rightarrow c \geq n/(2n-1)$$

$$\Rightarrow c \geq 1/(2-1/n)$$

$$\Rightarrow n_{0} = 1 \text{ and } c \geq 1$$

Example 2- Show that $T(n)=2T(\frac{n}{2})+n$ is $O(n\log n)$ by substitution method

Proof: To prove T(n) = O(nlogn), we need to prove that $T(n) \le cnlogn$. Assume that it is true for values smaller than n.

Induction Step: Assume it is true for $n=\frac{n}{2}$

i.e.,
$$T(\frac{n}{2}) \le 2.c. (\frac{n}{2}) \log (\frac{n}{2})$$
 is true

Now we have to show that it is true for n=n

i.e.
$$T(n) \le cnlogn$$

$$T(n) \le 2 \cdot T\left(\frac{n}{2}\right) + n$$

$$\le 2(c \cdot \left(\frac{n}{2}\right) \cdot \log\left(\frac{n}{2}\right)) + n$$

$$\le cn \log\left(\frac{n}{2}\right) + n \le cn \log n - cn \log 2 + n$$

$$\le cn \log n - cn + n \le cn \log n \ \forall c \ge 1$$
Thus $T(n) = O(nlogn)$

Example 2: Binary Search

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T(n) = c + T(n/2)
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- Guess: $T(n) = O(\log n)$
 - Induction goal: $T(n) \le d \log n$, for some d and $n \ge n_0$
 - Induction hypothesis: $T(n/2) \le d \log(n/2)$
- Proof of induction goal:

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

$$\leq d \log(n/2) + c$$

$$= d \log n - d + c = d \log n - (d - c)$$

$$\leq d \log n \quad \text{if: } d - c \geq 0, d \geq c$$

• What about n_0

The Iteration Method

Steps followed to solve any recurrence using iterating methods are:

- Expend the recurrence
- Express the expansion as a summation by plugging the recurrence back into itself until you see a pattern.
- Evaluate the summation by using the arithmetic or geometric summation formulae

Example 1:

$$T(n) = \begin{cases} n + T(n-1), & if n > 1 \\ 1, & if n \ge 1 \end{cases}$$

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

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

$$= n + (n-1) + (n-2) + T(n-3)$$

$$= \dots$$

$$= n + (n-1) + (n-2) + \dots + T(1), T(1) = 1, \text{ we can write the above equation as}$$

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

$$= n(n+1)/2 = (n^2+n)/2 \le n^2$$

$$= O(n^2)$$

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

•
$$T(n) = c + T(n/2)$$

 $= c + c + T(n/4)$
 $= c + c + c + T(n/8)$
Assume $n = 2^k$
 $T(n) = c + c + ... + c + T(1)$
 $= clogn + T(1)$
 $= O(logn)$

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

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

Example 3: T(n) = n + 2T(n/2)

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

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

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

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

$$= n + n + n + 8T(n/8)$$
... = $kn + 2^kT(n/2^k)$

$$= kn + 2^kT(1)$$

$$= nlogn + nT(1) = 0(nlgn)$$

Assume:
$$n = 2^k$$

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