Overview of Part 4

Theme:

Relating

Small problems or Big problems

* Proof techniques using

small soo Big relationships

- + Proof by smallest counter example
- + [Proof by induction]

* Problem solving techniques using

small mor Big relationships

- + Recursion
- + Divide 2 Conquer

* Running time of those techniques

+ Recurrence

L10-1

Overview of Part 4

L10:

* Smallest Counter Example

* Induction

LII:

Recursion

* Recurrence

L12:

* Divide & Conquer

* Recurrence

L13:

* Advanced Induction

strong Induction Example

P(n): n is power of prime humber or product of powers of prime number

prove: Vne N+ pcn) (A)

proof: Base case: 1 = 2°

pais true

Induction Hypothesis:

p(K) true for all k E Nt. K<n (f)

Induction: consider n

- 0 n = p = p' true
- (n=r·s r·seNt. <n
 - (4) => pers. pess true

=> p(n) true.

(A) is true.

Lio-3

Recapof LII

* Recursion:

Reduce problem of Size me into

Subproblems of Size h-1, ...

* Example: Towers of Hanoi

* Recurrences for running time of

recursive programs

$$M(n) = \begin{cases} 1 & n=1 \\ 2M(n-1)+1 & n>1 \end{cases} = 2^{n}-1$$

$$S(n) = \begin{cases} 0 & h=0 \\ 2S(n-1) & n>0 \end{cases} = 2^{n}$$

$$T(n) = \begin{cases} b & n=0 \\ rT(n-1) + a & n>0 \end{cases} = r^n b + a \frac{1-r^n}{1-r}$$

$$T(n) = \left\{ \begin{array}{l} \alpha & R^{(n)} \\ rT(n-1) + 5(n) & n > 0 \end{array} \right\} = r^{n} + \sum_{i=1}^{n} r^{n-i}g(i)$$

LIZ OVEY VIEW

* Divide & Conquer

Reduce problem of Size n

into subproblems of Size n/m

* Binary Search: Example

* Recurrences for Divide & Conquer programs running time of

$$T(n) = \begin{cases} 1 & n = 1 \\ T(\frac{n}{2}) + 1 & n > 1 \end{cases} = 1 + \log_2 n$$

$$T(n) = \begin{cases} \frac{1}{2} & n = 1 \\ 2T(\frac{n}{2}) + n & n > 1 \end{cases} = n + \log_2 n$$

$$T(n) = \begin{cases} 1 \\ 3T(\frac{h}{3}) + n \\ n \ge 3 \end{cases} = n + n \log_3 n$$

$$T(n) = \begin{cases} 1 & n = 1 \\ 4T(\frac{n}{2}) + n & n > 1 \end{cases} = 2n^2 - n$$

$$T(n) = \begin{cases} > 0 & n = 1 \\ a T(\frac{n}{2}) + n & n > 1 \end{cases}$$

$$T(n) = \begin{cases} G(n) & \alpha < 2 \\ G(n) & \alpha < 2 \\ G(n) & \alpha = 2 \end{cases}$$

$$G(n) & \alpha = 2$$

$$G(n) & \alpha > 2$$

* Lemma 4.3

Solving Recurrence for Binary Search

$$T(n) = \begin{cases} 1 & n = 1 \\ T(\frac{n}{2}) + 1 & n > 0 \end{cases}$$

Assume: n=z3 for some j

$$T(n) = T(\frac{n}{2}) + 1$$

$$=T(\frac{n}{2^3})+3$$

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

 $T(\frac{n}{2}) = T(\frac{n}{2}) + 1$

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

$$T(n) = 2T(\frac{n}{2}) + n$$

$$= 2(2T(\frac{n}{2^2}) + \frac{n}{2}) + n : T(\frac{n}{2}) = 2T(\frac{n}{4}) + \frac{n}{2}$$

$$= 2^2T(\frac{n}{2^2}) + 2n$$

$$= 2^2(2T(\frac{n}{2^3}) + \frac{n}{2^2}) + 2n : T(\frac{n}{2^2}) = 2T(\frac{n}{2^3}) + \frac{n}{2^2}$$

$$= 2^3T(\frac{n}{2^3}) + 3n$$

$$= 2^{i} T\left(\frac{n}{2^{i}}\right) + in$$

$$= 2^{j} T\left(\frac{n}{2^{j}}\right) + jn$$

$$= n T(1) + n \log_{2} n$$

$$= n + n \log_{2} n$$

$$T(n) = \begin{cases} 1 & n < 3 \\ 3T(\frac{n}{3}) + n & n > 3 \end{cases}$$

$$T(n) = 3T(\frac{n}{3}) + n$$

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

$$= 3^2 T(\frac{N}{3^2}) + 2n$$

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

$$=3^{3}T(\frac{N}{3^{3}})+3n$$

$$= 3^{i} T\left(\frac{n}{3^{i}}\right) + in$$

Assume n=33

$$= 3^{j} T\left(\frac{n}{3^{j}}\right) + jn$$

$$T(n) = \begin{cases} 1 & n=1 \\ 4T(\frac{n}{2}) + n & n>1 \end{cases}$$

$$T(\frac{1}{2}) = 4T(\frac{n}{2^2}) + \frac{n}{2}$$
$$= 4^2T(\frac{n}{2^2}) + \frac{4}{2}n + n$$

$$T(\frac{n}{2^2}) = 4T(\frac{n}{2^3}) + \frac{n}{2^2}$$

$$=4^{j}T(\frac{n}{z^{j}})+(\frac{t}{z})^{j-1}n+\dots+\frac{t}{z}n+n$$

$$=2^{2}i + 7(1) + n \sum_{i=0}^{j-1} (\frac{4}{2})^{i}$$

$$= n^2 + n \frac{2^{j-1}}{2-1}$$

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

Big-0 Examples

$$f(n) = 4 n^2$$

$$f(n) = 8n^2 + 2n - 3$$

$$\leq 10g(n)$$
, $n710 \Rightarrow f=0(g)$

$$f(n) = n^2/t + \sqrt{n} + \log n$$

$$\leq g(n)$$
, $n710 \Rightarrow partial$
 $f = o(g)$

In all cases, we can say: the running time is $O(n^2)$.

Convey essential info, & Simple.

use of hig-0, 0

 $4 n^2 \le 3n^2 + n \le 4n^2, n > 10$

If program has running time

3 n² +n,

we can simply say: running time is OCI

* nlogn & Ptonlogn & wonlogn, novo

Runningtime: Runningtime: Proposition

=> O(nlogn)

 $\frac{1}{2}$ $\frac{1}$

Running time: Pt lonlogn

 $\Rightarrow \Theta(n^2)$

Report Dominating term only

Lemm 4.3:
$$\sum_{i=0}^{n-1} Y^{i} = O(t(n))$$

claim:
$$t(n) = O(\sum_{i=1}^{h+1} r^i)$$
 (why?

$$\sum_{i=0}^{n-1} Y^{i} = o(f(n))$$

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

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

$$T(\frac{n}{2}) = T(\frac{n}{2^2}) + \frac{n}{2}$$

$$=T(\frac{n}{2^2})+\frac{n}{2}+n$$

$$T(\frac{n}{2^2}) = T(\frac{n}{2^3}) + \frac{n}{2^2}$$

$$= T(\frac{n}{2^2}) + \frac{n}{2^2} + \frac{n}{2} + n$$

Assume n= 21

$$= T(\frac{n}{2^{j}}) + \frac{n}{2^{j+1}} + \cdots + \frac{n}{2} + n$$

$$=1+0(n)$$

Theorem 4.4

$$= O(n)$$

Three recurrences

$$T(n) = 2T(\frac{n}{2}) + n = n + n \log_2 n = \Theta($$

$$T(n)=4T(\frac{n}{2})+n=2n^2-n=0$$

Lemma 4.7, case 3

Assume n=2 j.

Follow L13-7:

A STATE OF THE STA

$$= \alpha^{i} T(1) + n \sum_{i=0}^{j-1} \left(\frac{\alpha}{2}\right)^{i}$$

$$n\sum_{j=1}^{j-1}\left(\frac{a}{z}\right)^{j}=\Theta\left(n\left(\frac{a}{z}\right)^{j-1}\right)$$

$$=\Theta\left(n\frac{a^{j}}{z^{j}}\frac{z}{a}\right)=\Theta\left(n\frac{a^{\lfloor\theta\rfloor z^{n}}}{z^{\lfloor\theta\rfloor z^{n}}}\right)$$

$$T(n) = a^{\log_2 n} + \theta (a^{\log_2 n})$$