Lecture 11

Mathematical Analysis of Recursive Algorithms and Solving Recurrences: Recursive Tree Method, Master Theorem for Solving Recurrences.





The recursion-tree method

Convert the recurrence into a tree:

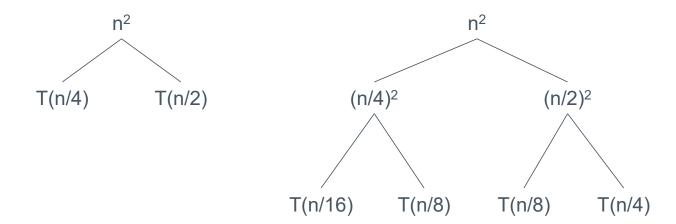
- Each node represents the cost incurred at various levels of recursion
- Sum up the costs of all levels

Used to "guess" a solution for the recurrence

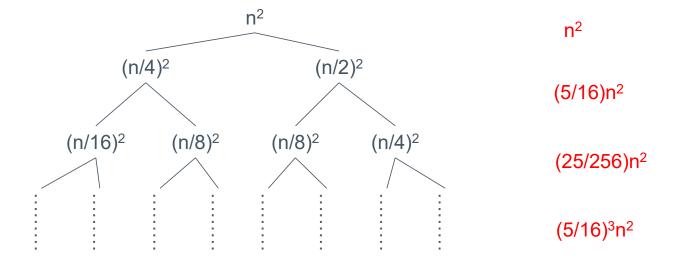


Recursion tree

- > Visualizing recursive tree method
- > eg. $T(n)=T(n/4)+T(n/2)+n^2$



Recursion tree (Cont !!!)



When the summation is infinite and |x| < 1, we have the infinite decreasing geometric series

$$\sum_{k=0}^{\infty} x^k = \frac{1}{1-x} \ . \tag{3.4}$$

$$n^{2} + \frac{5}{16}n^{2} + \frac{25}{256}n^{2} + \left(\frac{5}{16}\right)^{3}n^{2} + \dots$$

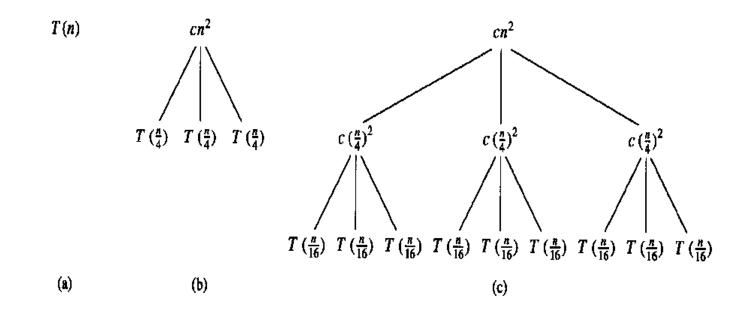
$$= n^{2} + \left(1 + \frac{5}{16} + \left(\frac{5}{16}\right)^{2} + \left(\frac{5}{16}\right)^{3} + \dots\right)$$

$$\cong n^{2} \cdot \frac{1}{1 - 5/16} = \Theta(n^{2})$$



Recursion-tree method (Cont !!!)

$$T(n) = 3T(\lfloor n/4 \rfloor) + \Theta(n^2)$$



Recursion-tree method (Cont !!!)

> Subproblem size for a node at depth i

$$\frac{n}{4^i}$$

> Total level of tree

$$\log_4 n + 1$$

> Number of nodes at depth i

$$3^i$$

> Cost of each node at depth i

$$c(\frac{n}{4^i})^2$$

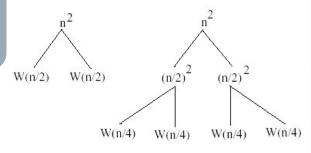
> Total cost at depth i

$$3^{i} c \left(\frac{n}{4^{i}}\right)^{2} = \left(\frac{3}{16}\right)^{i} c n^{2}$$

> Last level, depth $\log_4 n$, has $3^{\log_4 n} = n^{\log_4 3}$ nodes

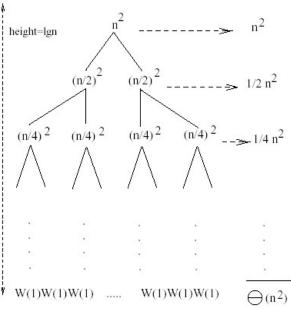


$$W(n) = 2W(n/2) + n^2$$



$$W(n/2)=2W(n/4)+(n/2)^{-2}$$

$$W(n/4)=2W(n/8)+(n/4)^{-2}$$



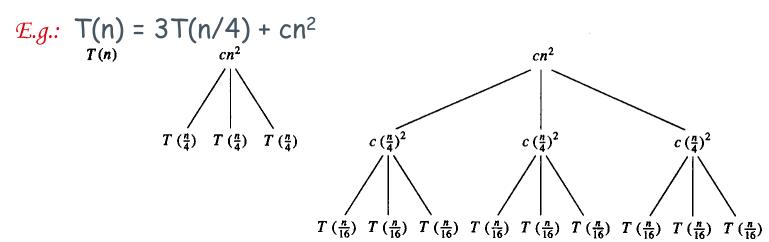
- > Subproblem size at level i is: $n/2^i$
- > Subproblem size hits 1 when $1 = n/2^i \Longrightarrow i = lgn$
- > Cost of the problem at level $i = (n/2^i)^2$ No. of nodes at level $i = 2^i$
- > Total cost:

$$W(n) = \sum_{i=0}^{\lg n-1} \frac{n^2}{2^i} + 2^{\lg n} W(1) = n^2 \sum_{i=0}^{\lg n-1} \left(\frac{1}{2}\right)^i + n \le n^2 \sum_{i=0}^{\infty} \left(\frac{1}{2}\right)^i + O(n) = n^2 \frac{1}{1 - \frac{1}{2}} + O(n) = 2n^2$$

$$\Rightarrow W(n) = O(n^2)$$

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Example



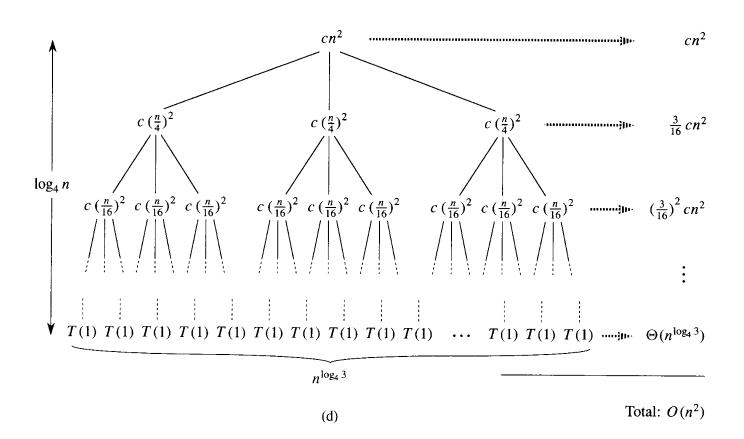
- Subproblem size at level i is: n/4ⁱ
- Subproblem size hits 1 when $1 = n/4^i \implies i = \log_4 n$
- Cost of a node at level $i = c(n/4^i)^2$
- Number of nodes at level $i = 3^i \Longrightarrow last level has 3^{log} = n^{log} 3^n nodes$
- Total cost:

$$T(n) = \sum_{i=0}^{\log_4 n - 1} \left(\frac{3}{16}\right)^i cn^2 + \Theta(n^{\log_4 3}) \le \sum_{i=0}^{\infty} \left(\frac{3}{16}\right)^i cn^2 + \Theta(n^{\log_4 3}) = \frac{1}{1 - \frac{3}{16}} cn^2 + \Theta(n^{\log_4 3}) = O(n^2)$$

$$\Rightarrow T(n) = O(n^2)$$



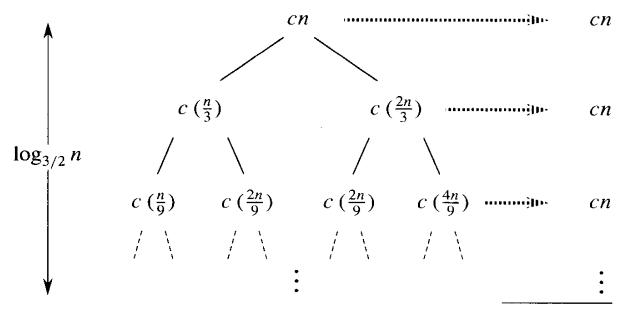
Recursion-tree method (Cont !!!)





Recursion-tree method((Cont !!!)

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



Total: $O(n \lg n)$



Explain the Master Method

- A utility method for analysing recurrence relations
- Useful in many cases for divide and conquer algorithms
- These recurrence relations are of the form:

$$T(n) = aT(n/b) + f(n)$$

with a >=1
and b >1

- n = the size of the current problem
- a = the number of subproblems in the recursion
- n/b = the size of each subproblem
- f(n) = the cost of the work that has to be done outside the recursive calls (cost of dividing + merging)



Explain the Master Method

The cases

There are 3 cases:

1. The running time is dominated by the cost at the leaves:

If
$$f(n) = O(n^{\log_b(a) \cdot \varepsilon})$$
, then $T(n) = O(n^{\log_b(a)})$

2. The running time is evenly distributed throughout the tree:

If
$$f(n) = \Theta(n^{\log_b(a)})$$
, then $T(n) = \Theta(n^{\log_b(a)}\log(n))$

3. The running time is dominated by the cost at the root:

If
$$f(n) = \Omega(n^{\log_b(a) + \varepsilon})$$
, then $T(n) = \Theta(f(n))$

If f(n) satisfies the regularity condition:

 $af(n/b) \le cf(n)$ where $c \le 1$ (this always holds for polynomials)

Because of this condition, the Master Method cannot solve every recurrence of the given form.

for an E > 0



How to apply the Master Method (step-by-step)

T(n) = aT(n/b) + f(n)

1. Extract a, b and f(n) from a given recurrence.

2. Determine n log_b(a)

3. Compare f(n) and $n^{\log_b(a)}$ asymptotically.

4. Determine the appropriate Master Method case and apply it.



Imagine that: T(n) = 2T(n/2) + n.

1. Extract;
$$a = 2$$
, $b = 2$ and $f(n) = n$.

2. Determine;
$$n^{\log_b(a)} = n^{\log_2(2)} = n^1 = n$$
.

3. Compare;
$$n = n$$

$$f(n) = n$$

4. Thus case 2; evenly distributed

Because
$$f(n) = \Theta(n)$$
,
 $T(n) = \Theta(n^{\log_b(a)}\log(n))$
 $= \Theta(n^{\log_b(a)}\log(n))$
 $= \Theta(n\log(n))$



+=

Imagine that: T(n) = 9T(n/3) + n.

1. Extract; a = 9, b = 3 and f(n) = n.

2. Determine; $n^{\log_b(a)} = n^{\log_3(9)} = n^2$.

3. Compare; $n = n^2$ f(n) = n

4. Thus case 1; (Express f(n) in terms of n

Because
$$f(n) = O(n^{2 \cdot \epsilon})$$
,
 $T(n) = O(n^{\log_b(a)}) = O(n^2)$.



Imagine that: $T(n) = 3T(n/4) + n\log(n)$.

1. Extract;
$$a = 3$$
, $b = 4$ and $f(n) = nlog(n)$.

2. Determine;
$$n^{\log_b(a)} = n^{\log_4(3)}$$
 where $\log_4(3) < 1$

3. Compare;
$$n = n \log_4(3)$$

$$f(n) = n \log(n)$$

4. Thus case 3, but we have to check the regularity condition!

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The following should be true: af(n/b) \le cf(n) where c \le 1 \le a(n/b)\log(n/b) \le cf(n) \le 3(n/4)\log(n/4) \le cf(n) \le 3/4 n\log(n/4) \le cf(n), this is true for c = 3/4, for example. So because f(n) = \Omega(n^{\log_4(3) + \epsilon}), f(n) = \Theta(f(n)) = \Theta(n\log(n))
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Analysis



Further Explanation

- > There are four methods to solve a recursive relation
 - Iterative
 - > In iterative method you will Convert the recurrence into a summation and try to bound it using known series.
 - Substitution
 - In substitution method, you will use guess or induction process to solve a recursive relation
 - Tree method
 - > In Tree method, you will form a tree and then sum up the values of nodes and also use guesses
 - Master Theorem:
 - > Only Specific problems can be solved in the form if recurrence relation is in the format like $T(n) = aT\left(\frac{n}{b}\right) + f(n)$ where $a \ge 1$ and b > 1



Master Theorem

> Let T(n) be a monotonically increasing function that satisfies

$$T(n) = a T(n/b) + f(n)$$

$$T(1) = c$$

where $a \ge 1$, $b \ge 2$ or b > 1, c > 0. If f(n) is $\Theta(n^d)$ where $d \ge 0$ then

$$\mathsf{T(n)} = \begin{cases} \Theta(n^d) & \text{if a < b^d} \\ \Theta(n^d \log n) & \text{if a = b^d} \\ \Theta(n^{\log_b a}) & \text{if a > b^d} \end{cases}$$



> Let $T(n) = T(n/2) + \frac{1}{2}n^2 + n$. What are the parameters?

$$a=1$$

$$b=2$$

$$d=2$$

$$T(n) = \begin{cases} \Theta(n^d) & \text{if } a < b^d \\ \Theta(n^d \log n) & \text{if } a > b^d \end{cases}$$

$$G(n^{\log_b a}) & \text{if } a > b^d \end{cases}$$

Therefore, which condition applies?

 $1 < 2^2$, case 1 applies

• We conclude that $T(n) \in \Theta(n^d) = \Theta(n^2)$



> Let $T(n) = 2 T(n/4) + \sqrt{n + 42}$. What are the parameters?

$$a = 2$$

$$d = 1/2$$

$$\mathsf{T(n)} = \begin{cases} \Theta(n^d) & \text{if a < b^d} \\ \Theta(n^d \log n) & \text{if a = b^d} \\ \Theta(n^{\log_b a}) & \text{if a > b^d} \end{cases}$$

Therefore, which condition applies?

$$2 = 4^{1/2}$$
, case 2 applies

$$T(n) \in \Theta(n^d \log n) = \Theta(\log n\sqrt{n})$$



> Let T(n) = 3 T(n/2) + 3/4n + 1. What are the parameters?

$$a = 3$$
 $b = 2$
 $d = 1$
T(n

$$\mathsf{T(n)} = \begin{cases} \Theta(n^d) & \text{if a < b^d} \\ \Theta(n^d \log n) & \text{if a = b^d} \\ \Theta(n^{\log_b a}) & \text{if a > b^d} \end{cases}$$

Therefore, which condition applies?

 $3 > 2^1$, case 3 applies

We conclude that

$$T(n) \in \Theta(n^{\log_b a}) = \Theta(n^{\log_2 3})$$

• Note that $\log_2 3 \approx 1.584...$, can we say that $T(n) \in \Theta$ $(n^{1.584})$

No, because $log_2 3 \approx 1.5849...$ and $n^{1.584} \notin \Theta$ $(n^{1.5849})$



> Let $T(n) = 2T(n/2) + n \log n$. What are the parameters?

$$a = 2$$

$$b = 2$$

$$d = 1$$

$$T(n) = \begin{cases} \Theta(n^d) & \text{if } a < b^d \\ \Theta(n^d \log n) & \text{if } a = b^d \\ \Theta(n^{\log_b a}) & \text{if } a > b^d \end{cases}$$

Therefore, which condition applies?

$$2 = 2^{1}$$
, case 2 applies

$$T(n) = \Theta (n^1 \log n) = \Theta (n \log n)$$



> Let $T(n) = T(n/3) + n \log n$. What are the parameters?

$$a = 1$$

$$b = 3$$

$$d = 1$$

$$T(n) = \begin{cases} \Theta(n^d) & \text{if } a < b^d \\ \Theta(n^d \log n) & \text{if } a = b^d \\ \Theta(n^{\log_b a}) & \text{if } a > b^d \end{cases}$$

Therefore, which condition applies?

$$1 < 3^1$$
, case 1 applies

$$T(n) = \Theta(n^1) = \Theta(n)$$



> Let T(n) = 4T(n/2) + n. What are the parameters?

$$a = 4$$

$$b = 2$$

$$d = 1$$

$$T(n) = \begin{cases} \Theta(n^d) & \text{if } a < b^d \\ \Theta(n^d \log n) & \text{if } a = b^d \\ \Theta(n^{\log_b a}) & \text{if } a > b^d \end{cases}$$

Therefore, which condition applies?

$$4 > 2^{1}$$
, case 3 applies

$$T(n) = \Theta (n^{\log_b a}) = \Theta (n^{\log_2 4}) = \Theta(n^2)$$



> Let $T(n) = 8T(n/2) + n^2$. What are the parameters?

$$a = 8$$

$$b = 2$$

$$d = 2$$

$$T(n) = \begin{cases} \Theta(n^d) & \text{if } a < b^d \\ \Theta(n^d \log n) & \text{if } a = b^d \\ \Theta(n^{\log_b a}) & \text{if } a > b^d \end{cases}$$

Therefore, which condition applies?

$$8 > 2^2$$
, case 3 applies

$$T(n) = \Theta (n^{\log_2 8}) = \Theta (n^{\log_2 2^3}) = \Theta (n^3)$$



> Let $T(n) = 9T(n/3) + n^3$. What are the parameters?

$$a = 9$$

$$b = 3$$

$$d = 3$$

$$T(n) = \begin{cases} \Theta(n^d) & \text{if } a < b^d \\ \Theta(n^d \log n) & \text{if } a = b^d \\ \Theta(n^{\log_b a}) & \text{if } a > b^d \end{cases}$$

Therefore, which condition applies?

$$9 = 3^3$$
, case 2 applies

$$T(n) = \mathbf{\Theta} (n^3 \log n) = \mathbf{\Theta} (n^3 \log n)$$



> Let T(n) = T(n/2) + 1. What are the parameters?

$$a = 1$$

$$b = 2$$

$$d = 0$$

$$T(n) = \begin{cases} \Theta(n^d) & \text{if } a < b^d \\ \Theta(n^d \log n) & \text{if } a = b^d \\ \Theta(n^{\log_b a}) & \text{if } a > b^d \end{cases}$$

Therefore, which condition applies?

$$1 = 2^0$$
, case 2 applies

$$T(n) = \mathbf{\Theta} (n^0 \log n) = \mathbf{\Theta} (\log n)$$



Recurrence Relation

$$T(n) = \begin{cases} b & \text{if } n < 2\\ 2T(n/2) + bn & \text{if } n \ge 2 \end{cases}$$



Iterative Substitution

> In the iterative substitution, or "plug-and-chug," technique, we iteratively apply the recurrence equation to itself and see if we can find a pattern:

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

$$= 2(2T(n/2^{2})) + b(n/2)) + bn$$

$$= 2^{2}T(n/2^{2}) + 2bn$$

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

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

$$= ...$$

$$= 2^{i}T(n/2^{i}) + ibn$$

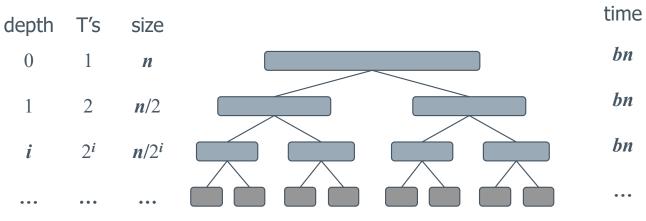
- Note that base, T(n)=b, case occurs when $2^i=n$. That is, $i=\log n$. So, $T(n)=bn+bn\log n$
- \rightarrow Thus, T(n) is $O(n \log n)$.



The Recursion Tree

> Draw the recursion tree for the recurrence relation and look for a pattern:

$$T(n) = \begin{cases} b & \text{if } n < 2\\ 2T(n/2) + bn & \text{if } n \ge 2 \end{cases}$$



 $Total\ time = bn + bn \log n$

Master Theorem:

> Let T(n) = 2T(n/2) + bnlogn. What are the parameters?

Therefore, which condition applies?

$$2 = 2^{1}$$
, case 2 applies

$$T(n) = \Theta (n^1 \log n) = \Theta (n \log n)$$

Summary

 \rightarrow Let T(n) be a monotonically increasing function that satisfies

$$T(n) = a T(n/b) + f(n)$$

$$T(1) = c$$

where $a \ge 1$, $b \ge 2$, c > 0. If f(n) is $\Theta(n^d)$ where $d \ge 0$ then

$$\mathsf{T(n)} = \begin{cases} \Theta(n^d) & \text{if a < b^d} \\ \Theta(n^d \log n) & \text{if a = b^d} \\ \Theta(n^{\log_b a}) & \text{if a > b^d} \end{cases}$$

Thank You!!!

Have a good day

