

CSE408 Recurrence equations

Lecture #12

Substitution method



The most general method:

- 1. Guess the form of the solution.
- 2. Verify by induction.
- 3. Solve for constants.

Substitution method



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- 1. Guess the form of the solution.
- 2. Verify by induction.
- 3. Solve for constants.

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

- [Assume that $T(1) = \Theta(1)$.]
- Guess $O(n^3)$. (Prove O and Ω separately.)
- Assume that $T(k) \le ck^3$ for k < n.
- Prove $T(n) \leq cn^3$ by induction.

Example of substitution



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

 $\leq 4c(n/2)^3 + n$
 $= (c/2)n^3 + n$
 $= cn^3 - ((c/2)n^3 - n) \leftarrow desired - residual$
 $\leq cn^3 \leftarrow desired$
whenever $(c/2)n^3 - n \geq 0$, for example, if $c \geq 2$ and $n \geq 1$.
 $residual$

Example (continued)



- We must also handle the initial conditions, that is, ground the induction with base cases.
- **Base:** $T(n) = \Theta(1)$ for all $n < n_0$, where n_0 is a suitable constant.
- For $1 \le n < n_0$, we have " $\Theta(1)$ " $\le cn^3$, if we pick c big enough.

Example (continued)



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This bound is not tight!

Recursion-tree method



- A recursion tree models the costs (time) of a recursive execution of an algorithm.
- The recursion-tree method can be unreliable, just like any method that uses ellipses (...).
- The recursion-tree method promotes intuition, however.
- The recursion tree method is good for generating guesses for the substitution method.



Solve
$$T(n) = T(n/4) + T(n/2) + n^2$$
:



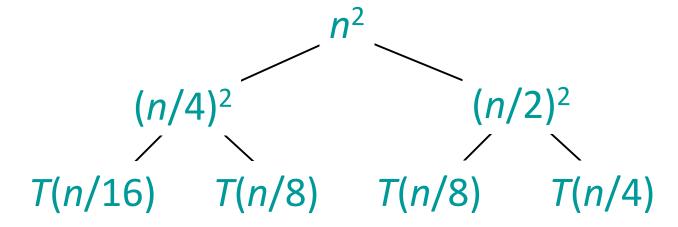
Solve
$$T(n) = T(n/4) + T(n/2) + n^2$$
:
$$T(n)$$



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:
$$T(n/4) \qquad T(n/2)$$



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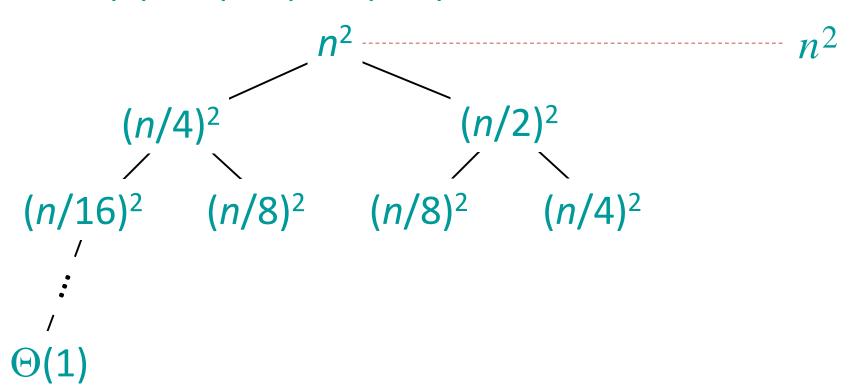


Solve
$$T(n) = T(n/4) + T(n/2) + n^2$$
:

```
(n/4)^2 (n/2)^2 (n/16)^2 (n/8)^2 (n/8)^2 (n/4)^2 \vdots
```

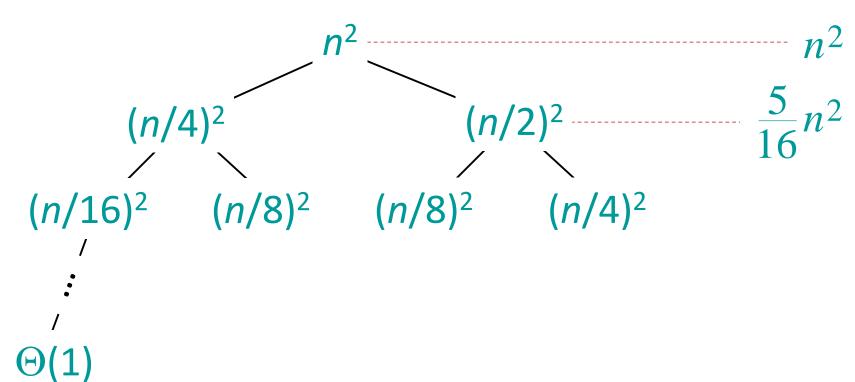


```
Solve T(n) = T(n/4) + T(n/2) + n^2:
```



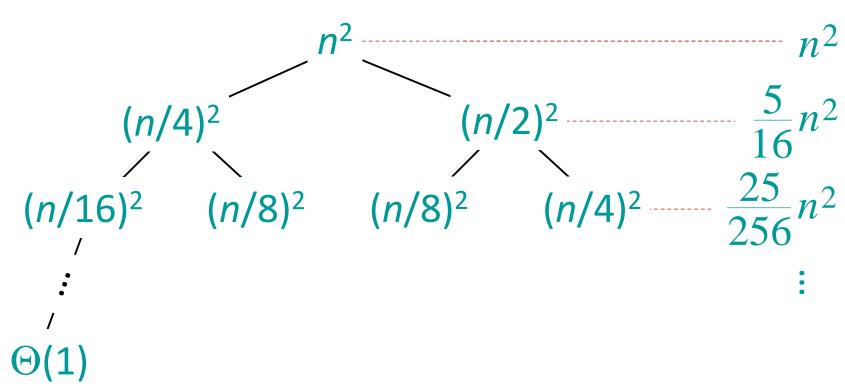


Solve $T(n) = T(n/4) + T(n/2) + n^2$:





Solve
$$T(n) = T(n/4) + T(n/2) + n^2$$
:





Solve $T(n) = T(n/4) + T(n/2) + n^2$:

$$(n/4)^{2} \qquad (n/2)^{2} \qquad \frac{5}{16}n^{2}$$

$$(n/16)^{2} \qquad (n/8)^{2} \qquad (n/4)^{2} \qquad \frac{25}{256}n^{2}$$

$$\vdots \qquad \vdots \qquad \vdots$$

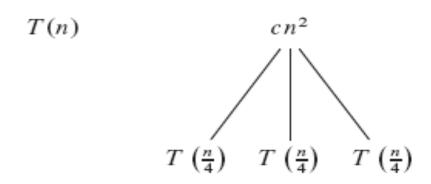
$$\Theta(1) \qquad \text{Total} = n^{2}\left(1 + \frac{5}{16} + \left(\frac{5}{16}\right)^{2} + \left(\frac{5}{16}\right)^{3} + \cdots\right)$$

$$= \Theta(n^{2}) \qquad \text{geometric series} \quad \bullet$$

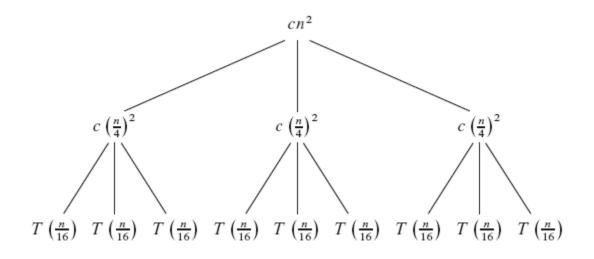


$$T(n) = 3T(n/4) + cn^2.$$

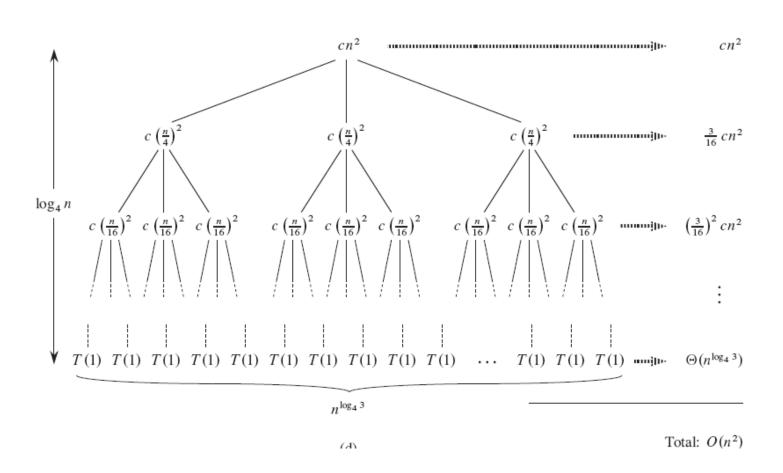














Because subproblem sizes decrease by a factor of 4 each time we go down one level, we eventually must reach a boundary condition. How far from the root do we reach one? The subproblem size for a node at depth i is $n/4^i$. Thus, the subproblem size hits n = 1 when $n/4^i = 1$ or, equivalently, when $i = \log_4 n$. Thus, the tree has $\log_4 n + 1$ levels (at depths $0, 1, 2, ..., \log_4 n$).



Next we determine the cost at each level of the tree. Each level has three times more nodes than the level above, and so the number of nodes at depth i is 3^{i} .



Because subproblem sizes reduce by a factor of 4 for each level we go down from the root, each node at depth i, for $i = 0, 1, 2, ..., \log_4 n - 1$, has a cost of $c(n/4^i)^2$. Multiplying, we see that the total cost over all nodes at depth i, for $i = 0, 1, 2, ..., \log_4 n - 1$, is $3^i c(n/4^i)^2 = (3/16)^i cn^2$. The bottom level, at depth $\log_4 n$, has $3^{\log_4 n} = n^{\log_4 3}$ nodes, each contributing cost T(1), for a total cost of $n^{\log_4 3}T(1)$, which is $\Theta(n^{\log_4 3})$, since we assume that T(1) is a constant.



$$T(n) = cn^{2} + \frac{3}{16}cn^{2} + \left(\frac{3}{16}\right)^{2}cn^{2} + \dots + \left(\frac{3}{16}\right)^{\log_{4}n - 1}cn^{2} + \Theta(n^{\log_{4}3})$$

$$= \sum_{i=0}^{\log_{4}n - 1} \left(\frac{3}{16}\right)^{i}cn^{2} + \Theta(n^{\log_{4}3})$$

$$= \frac{(3/16)^{\log_{4}n} - 1}{(3/16) - 1}cn^{2} + \Theta(n^{\log_{4}3}) \qquad \text{(by equation (A.5))}.$$



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

$$< \sum_{i=0}^{\infty} \left(\frac{3}{16}\right)^i cn^2 + \Theta(n^{\log_4 3})$$

$$= \frac{1}{1 - (3/16)} cn^2 + \Theta(n^{\log_4 3})$$

$$= \frac{16}{13} cn^2 + \Theta(n^{\log_4 3})$$

$$= O(n^2).$$

The master method



The master method applies to recurrences of the form

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

where $a \ge 1$, b > 1, and f is asymptotically positive.

Three common cases



Compare f(n) with $n^{\log_b a}$:

- 1. $f(n) = O(n^{\log_b a \varepsilon})$ for some constant $\varepsilon > 0$.
 - f(n) grows polynomially slower than $n^{\log b^{\alpha}}$ (by an n^{ϵ} factor).

Solution: $T(n) = \Theta(n^{\log_b a})$.

Three common cases



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Solution: T(n) = \Theta(n^{\log_b a}).
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- 2. $f(n) = \Theta(n^{\log_b a} \lg^k n)$ for some constant $k \ge 0$.
 - f(n) and $n^{\log_b a}$ grow at similar rates.

Solution:
$$T(n) = \Theta(n^{\log_b a} \lg^{k+1} n)$$
.

Three common cases



Compare f(n) with $n^{\log_b a}$:

- 3. $f(n) = \Omega(n^{\log_b a + \varepsilon})$ for some constant $\varepsilon > 0$.
 - f(n) grows polynomially faster than $n^{\log_b a}$ (by an n^{ϵ} factor),

and f(n) satisfies the regularity condition that $af(n/b) \le cf(n)$ for some constant c < 1.

Solution: $T(n) = \Theta(f(n))$.



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

 $a = 4, b = 2 \Rightarrow n^{\log_b a} = n^2; f(n) = n.$
Case 1: $f(n) = O(n^{2-\epsilon})$ for $\epsilon = 1$.
 $\therefore T(n) = \Theta(n^2).$



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Ex.
$$T(n) = 4T(n/2) + n^2$$

 $a = 4, b = 2 \Rightarrow n^{\log_b a} = n^2; f(n) = n^2.$
Case 2: $f(n) = \Theta(n^2 \lg^0 n)$, that is, $k = 0$.
 $T(n) = \Theta(n^2 \lg n)$.



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

 $a = 4, b = 2 \Rightarrow n^{\log_b a} = n^2; f(n) = n^3.$
Case 3: $f(n) = \Omega(n^{2+\epsilon})$ for $\epsilon = 1$
 $and \ 4(n/2)^3 \le cn^3$ (reg. cond.) for $c = 1/2$.
 $\therefore T(n) = \Theta(n^3).$



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

 $a = 4, b = 2 \Rightarrow n^{\log_b a} = n^2; f(n) = n^3.$
Case 3: $f(n) = \Omega(n^{2+\epsilon})$ for $\epsilon = 1$
and $4(n/2)^3 \le cn^3$ (reg. cond.) for $c = 1/2$.
 $\therefore T(n) = \Theta(n^3).$

Ex.
$$T(n) = 4T(n/2) + n^2/\lg n$$

 $a = 4, b = 2 \Rightarrow n^{\log_b a} = n^2; f(n) = n^2/\lg n.$
Master method does not apply. In particular, for every constant $\varepsilon > 0$, we have $n^{\varepsilon} = \omega(\lg n)$.



Thank You !!!