

# Lecture 3: The Master Theorem

## 1 Quick Review

- Recall, a divide-and-conquer algorithm recursively breaks up a problem of size  $n$  in smaller sub-problems such that:
  - There are exactly  $a$  sub-problems.
  - Each sub-problem has size at most  $\frac{1}{b} \cdot n$
  - Once solved, the solutions to the sub-problems can be combined to produce a solution to the original problem in time  $O(n^d)$
- So the run-time of a divide and conquer algorithm satisfies the recurrence:

$$T(n) = a \cdot T\left(\frac{n}{b}\right) + O(n^d)$$

- Examples: MergeSort and Binary Search

## 2 The Master Theorem

- **The Master Theorem:** If  $T(n) = a \cdot T\left(\frac{n}{b}\right) + O(n^d)$  for constants  $a > 0$ ,  $b > 1$ , and  $d \geq 0$  then:

$$T(n) = \begin{cases} O(n^d) & \text{if } a < b^d \text{ [Case I]} \\ O(n^d \cdot \log n) & \text{if } a = b^d \text{ [Case II]} \\ O(n^{\log_b a}) & \text{if } a > b^d \text{ [Case III]} \end{cases}$$

- Sanity Check: What does this give for MergeSort and Binary Search?

	$a$	$b$	$d$	Case	Runtime
MergeSort	2	2	1	II	$O(n \cdot \log n)$
Binary Search	1	2	0	II	$O(\log n)$

### 3 Proof of The Master Theorem

- Fact One:

Fact 1.  $\sum_{k=0}^{\ell} \tau^k = \frac{1 - \tau^{\ell+1}}{1 - \tau}$  for any  $\tau \neq 1$ .

Proof.

- We have:

$$\begin{aligned}
 (1 - \tau) \cdot \sum_{k=0}^{\ell} \tau^k &= \sum_{k=0}^{\ell} \tau^k - \sum_{k=1}^{\ell+1} \tau^k \\
 &= \tau^0 - \tau^{\ell+1} \\
 &= 1 - \tau^{\ell+1}
 \end{aligned}$$

- Dividing both sides by  $(1 - \tau)$  gives the result.

- Fact Two:

Fact 2.  $x^{\log_b y} = y^{\log_b x}$  for any base  $b$ .

Proof.

$\log_b z^p = p \cdot \log_b z$

- Observe that, by the power rule of logarithms, we have:
 
$$\log_b x \cdot \log_b y = \log_b (y^{\log_b x})$$
- Similarly:
 
$$\log_b x \cdot \log_b y = \log_b (x^{\log_b y})$$
- Putting this together gives
 
$$\log_b (y^{\log_b x}) = \log_b (x^{\log_b y})$$

$$\Rightarrow x^{\log_b y} = y^{\log_b x}$$

- Proof of the Master Theorem

Proof.

- We may assume  $n$  is a power of  $b$ :  $n = b^\ell$
- So we have:
 
$$T(n) = n^d + a \cdot \left(\frac{n}{b}\right)^d + a^2 \cdot \left(\frac{n}{b^2}\right)^d + \dots + a^\ell \cdot \left(\frac{n}{b^\ell}\right)^d$$

$$= n^d \cdot \left(1 + a \cdot \left(\frac{1}{b}\right)^d + a^2 \cdot \left(\frac{1}{b^2}\right)^d + \dots + a^\ell \cdot \left(\frac{1}{b^\ell}\right)^d\right)$$

$$= n^d \cdot \left(1 + \frac{a}{b^d} + \left(\frac{a}{b^d}\right)^2 + \dots + \left(\frac{a}{b^d}\right)^\ell\right)$$

Proof [cont.]  $T(n) = n^d \cdot \left( 1 + \frac{a}{b^d} + \left( \frac{a}{b^d} \right)^2 + \dots + \left( \frac{a}{b^d} \right)^\ell \right)$

Case I:  $\frac{a}{b^d} < 1$

- Set  $\tau = \frac{a}{b^d}$

- Then:  $T(n) = n^d \cdot \sum_{k=0}^{\ell} \tau^k$

- Applying Fact 1, we know that:

$$\sum_{k=0}^{\ell} \tau^k = \frac{1 - \tau^{\ell+1}}{1 - \tau} \leq \frac{1}{1 - \tau} = O(1)$$

As  $a$ ,  $b$ , and  $d$  are constants so is  $1 - \tau$ .

- Therefore:

$$T(n) \leq n^d \cdot \frac{1}{1 - \frac{a}{b^d}} = n^d \cdot \frac{b^d}{b^d - a} = O(n^d)$$

Proof [cont.]  $T(n) = n^d \cdot \left( 1 + \frac{a}{b^d} + \left( \frac{a}{b^d} \right)^2 + \dots + \left( \frac{a}{b^d} \right)^\ell \right)$

Case II:  $\frac{a}{b^d} = 1$

- Then:  $T(n) = n^d \cdot (\ell + 1)$

- But  $n = b^\ell$  so  $\ell = \log_b n$ .

- As  $b$  is a constant greater than one, this gives:  $T(n) = O(n^d \cdot \log n)$

Case III:  $\frac{a}{b^d} > 1$

- Again set  $\tau = \frac{a}{b^d}$

- Then:  $T(n) = n^d \cdot \sum_{k=0}^{\ell} \tau^k$

- Applying Fact 1 gives:

$$\sum_{k=0}^{\ell} \tau^k = \frac{\tau^{\ell+1} - 1}{\tau - 1} \leq \frac{\tau^{\ell+1}}{\tau - 1} = O(\tau^{\ell+1}) = O(\tau^\ell)$$

As  $a$ ,  $b$ , and  $d$  are constants so is  $\tau$ .

Proof [cont.]

▪ Thus:  $T(n) = O(n^d \cdot \tau^\ell)$

▪ Observe that:

$$n^d \cdot \tau^\ell = n^d \cdot \left(\frac{a}{b^d}\right)^\ell$$

$$= \left(\frac{n}{b^\ell}\right)^d \cdot a^\ell$$

$$= 1 \cdot a^\ell \quad \text{--- As } n=b^\ell.$$

$$= a^{\log_b n}$$

$$= n^{\log_b a} \quad \text{--- By Fact 2.}$$

▪ This gives the final case:

$$T(n) = O(n^{\log_b a})$$

- Specifically, what matters is **not** the statement of the Master Theorem but the **ideas** underlying its proof.

- First, if we understand the proof then we can easily reconstruct the theorem.
- Second, if we understand the proof then we can easily apply the method to a much broader class of problems. For example:

→ The sub-problems have different sizes.

e.g. The Deterministic Selection Algorithm.

$$T(n) \leq T\left(\frac{7n}{10}\right) + T\left(\frac{n}{5}\right) + O(n)$$

→ The combination function is not of the form  $f(n) = n^d$ .

e.g. Euclid's Greatest Common Divisor Algorithm.

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

→ The parameters a, b, and d are not constants.

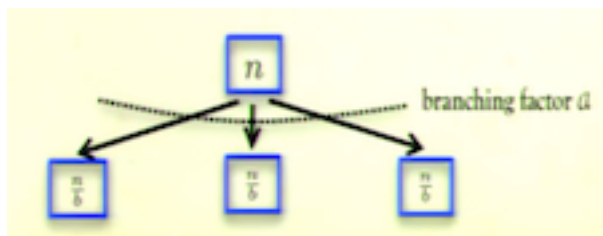
$$\text{e.g. } T(n) = \sqrt{n} \cdot T(\sqrt{n}) + O(n^{\frac{1}{\log \log n}})$$

## 4 The Recursion Tree Method

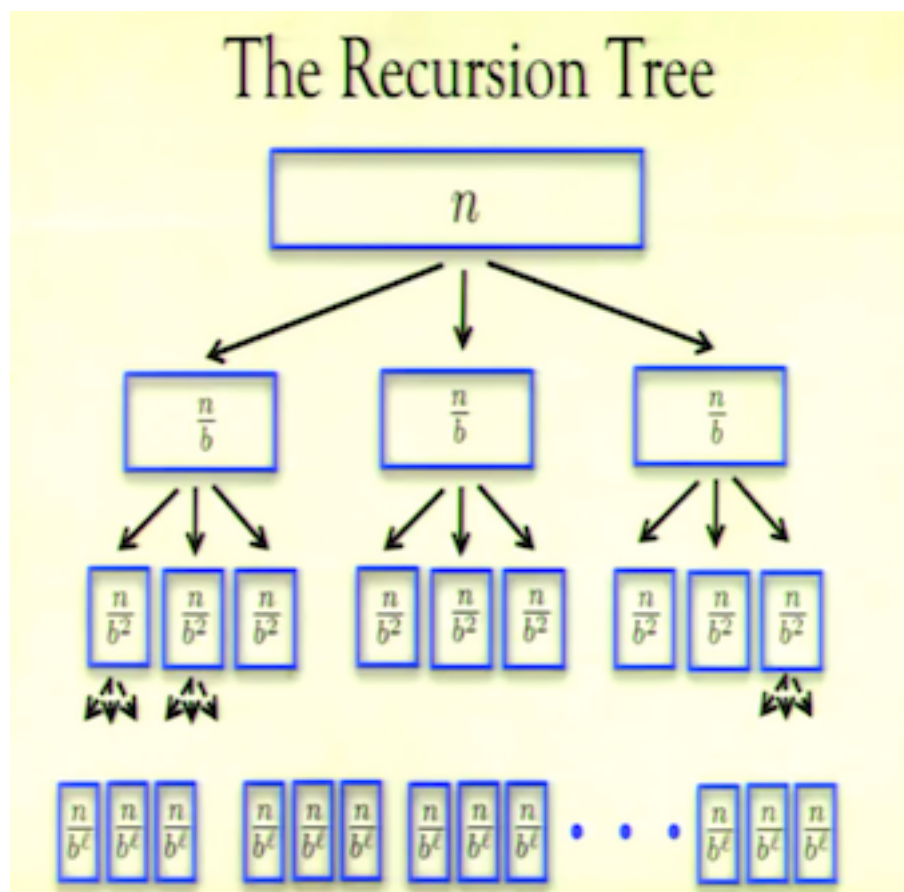
- The Master Theorem is a special case the the **recursion tree method**.
- Specifically, we model the *divide and conquer* recursive formula by a tree:

$$T(n) = a \cdot T\left(\frac{n}{b}\right) + O(n^d)$$

- The **root** node of the trees has a label n.
- The root has a children each with label  $\frac{n}{b}$ .

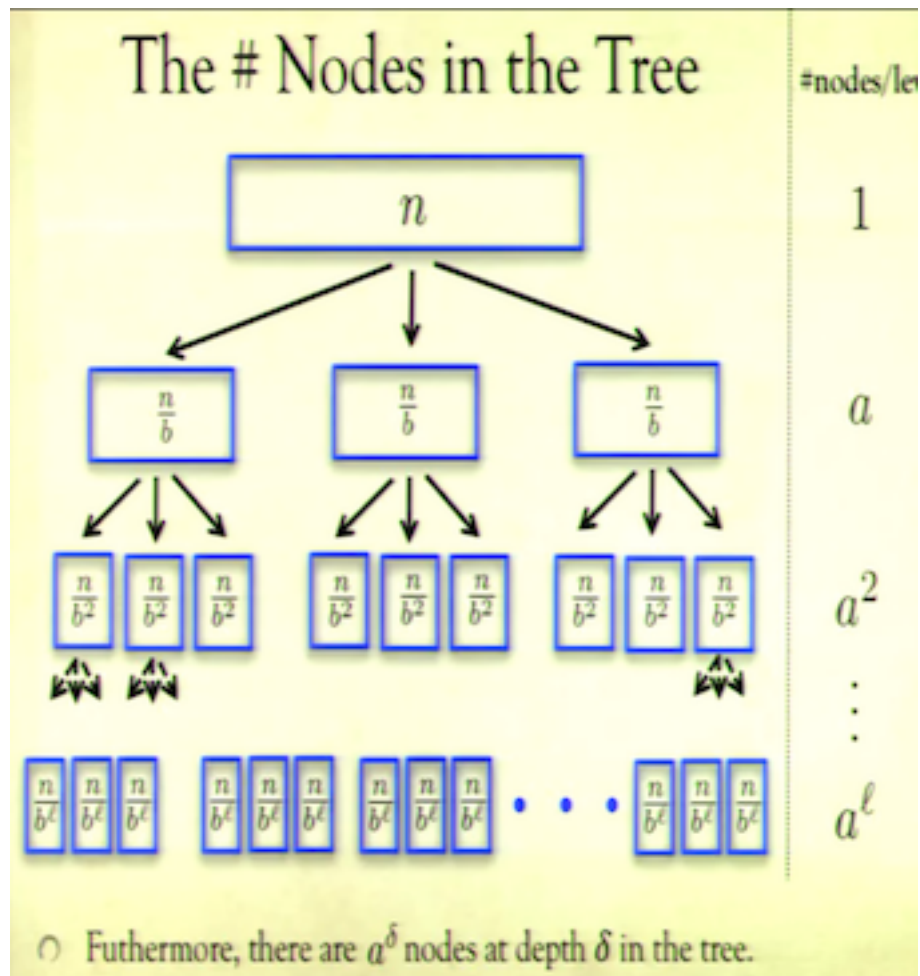


- This pattern then repeats at the children, then grandchildren, etc.

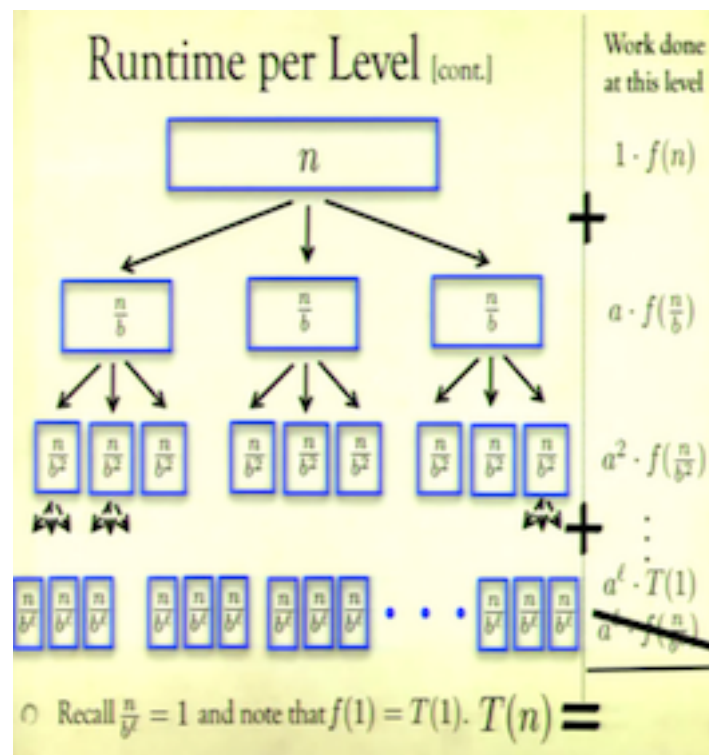




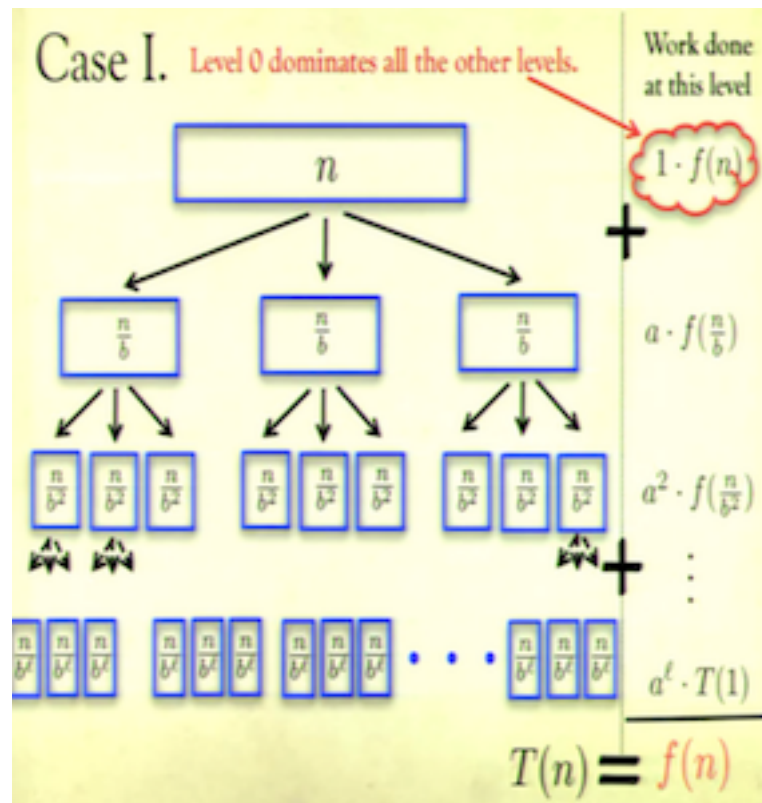
- This process stops at the **leaves** (base cases) which have label  $\frac{n}{b^l} = 1$ . (As  $n = b^l$ )



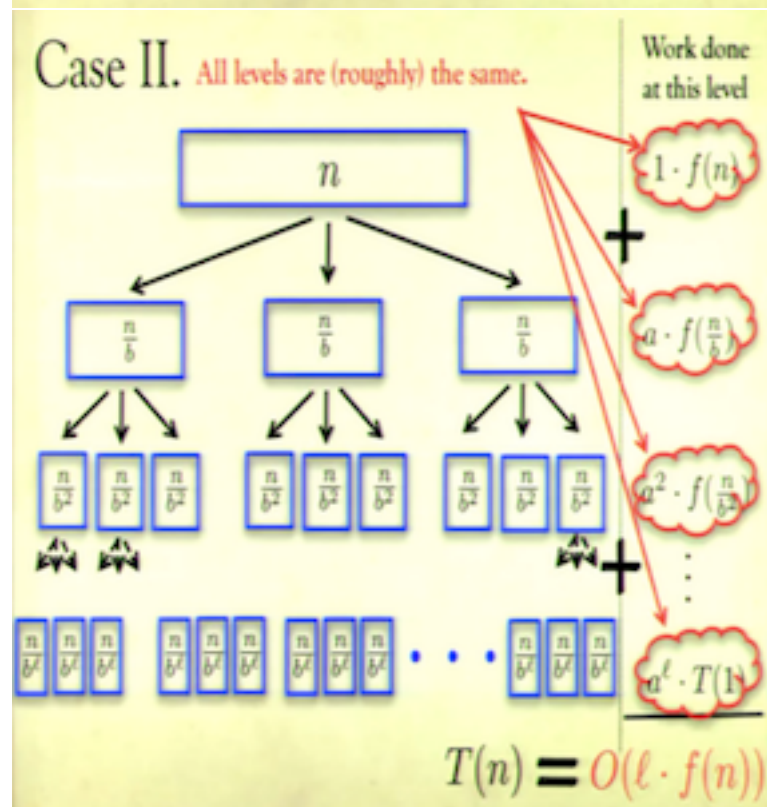
- How much time do we spend at each level?



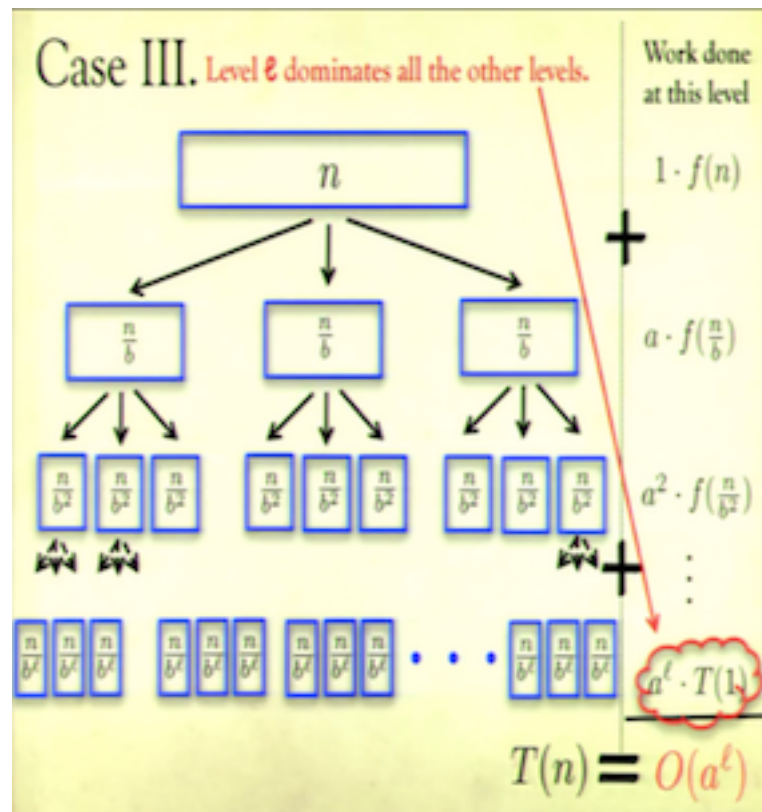
Case I. Level 0 dominates all the other levels.



Case II. All levels are (roughly) the same.







- This gives us the proof of the Master Theorem:

