# CS60020: Foundations of Algorithm Design and Machine Learning

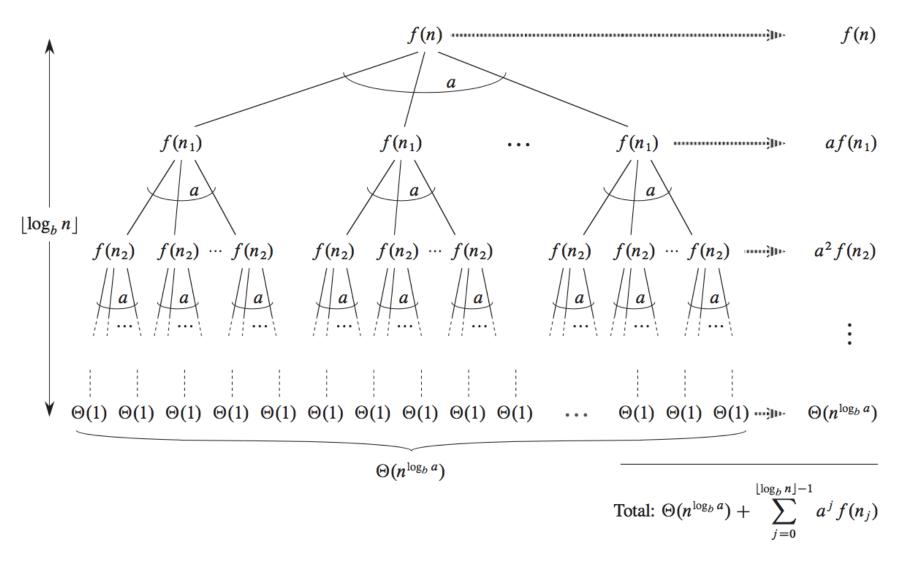
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## Master theorem

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

**CASE 1:**  $f(n) = O(n^{\log_b a - \varepsilon})$ , constant  $\varepsilon > 0$  $\Rightarrow T(n) = \Theta(n^{\log_b a})$ . **CASE 2:**  $f(n) = \Theta(n^{\log_b a})$  $\Rightarrow T(n) = \Theta(n^{\log_b a} \lg n)$ . **CASE 3:**  $f(n) = \Omega(n^{\log_b a + \varepsilon})$ , constant  $\varepsilon > 0$ , and regularity condition  $\Rightarrow$   $T(n) = \Theta(f(n))$ .



#### Lemma 4.3

Let  $a \ge 1$  and b > 1 be constants, and let f(n) be a nonnegative function defined on exact powers of b. A function g(n) defined over exact powers of b by

$$g(n) = \sum_{j=0}^{\log_b n-1} a^j f(n/b^j)$$
(4.22)

has the following asymptotic bounds for exact powers of b:

- 1. If  $f(n) = O(n^{\log_b a \epsilon})$  for some constant  $\epsilon > 0$ , then  $g(n) = O(n^{\log_b a})$ .
- 2. If  $f(n) = \Theta(n^{\log_b a})$ , then  $g(n) = \Theta(n^{\log_b a} \lg n)$ .
- 3. If  $af(n/b) \le cf(n)$  for some constant c < 1 and for all sufficiently large n, then  $g(n) = \Theta(f(n))$ .

• Case 1:

$$\sum_{j=0}^{\log_b n-1} a^j \left(\frac{n}{b^j}\right)^{\log_b a-\epsilon} = n^{\log_b a-\epsilon} \sum_{j=0}^{\log_b n-1} \left(\frac{ab^\epsilon}{b^{\log_b a}}\right)^j$$
$$= n^{\log_b a-\epsilon} \sum_{j=0}^{\log_b n-1} (b^\epsilon)^j$$
$$= n^{\log_b a-\epsilon} \left(\frac{b^{\epsilon \log_b n}-1}{b^{\epsilon}-1}\right)$$

• Case 2:

$$\sum_{j=0}^{\log_b n-1} a^j \left(\frac{n}{b^j}\right)^{\log_b a} = n^{\log_b a} \sum_{j=0}^{\log_b n-1} \left(\frac{a}{b^{\log_b a}}\right)^j$$
$$= n^{\log_b a} \sum_{j=0}^{\log_b n-1} 1$$
$$= n^{\log_b a} \log_b n.$$

• Case 3:

$$g(n) = \sum_{j=0}^{\log_b n-1} a^j f(n/b^j)$$
  

$$\leq \sum_{j=0}^{\log_b n-1} c^j f(n) + O(1)$$
  

$$\leq f(n) \sum_{j=0}^{\infty} c^j + O(1)$$
  

$$= f(n) \left(\frac{1}{1-c}\right) + O(1)$$
  

$$= O(f(n)),$$