# COMP 3711 Design and Analysis of Algorithms

## Master Theorem

### Divide-and-Conquer Recurrence Examples

#### Major examples so far

- Maximum Contiguous Subarray & Mergesort
  - Both satisfied T(n) = 2T(n/2) + O(n)
  - $T(n) = O(n \log n)$
- First version of Integer Multiplication
  - T(n) = 4T(n/2) + O(n)
  - $T(n) = O(n^2)$
- Karatsuba Multiplication
  - T(n) = 3T(n/2) + O(n)
  - $T(n) = O(n^{\log_2 3}) = O(n^{1.58...})$

#### The Master Theorem

Tool for directly (i.e., without expansion or recurrence tree) solving recurrences of form

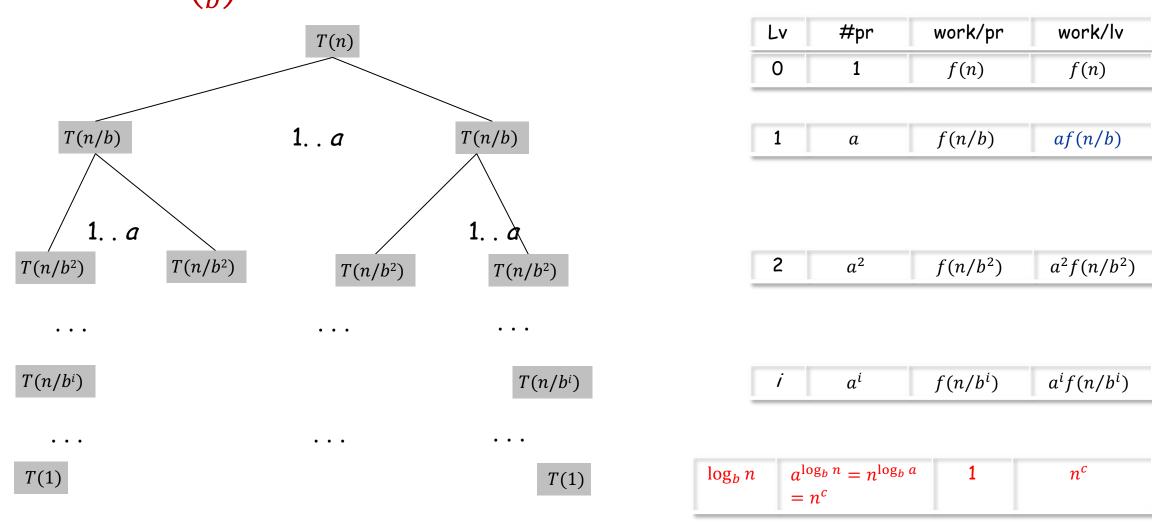
$$T(n) = aT(n/b) + f(n)$$

#### where

- $a \ge 1$  and b > 1 are constants and
- f(n) is a (asymptotically) positive polynomial function.
- Initial conditions T(1), T(2), ..., T(k) for some k don't contribute to asymptotic growth
- n/b could be either  $\lfloor n/b \rfloor$  or  $\lfloor n/b \rfloor$

#### Visualization for Master Theorem

$$T(n) = aT\left(\frac{n}{b}\right) + f(n), \quad a \ge 1$$
 and  $b > 1$  are constants,  $c = \log_b a$ 



## The Master Theorem (for equalities)

$$T(n) = aT(n/b) + f(n), \qquad c = \log_b a$$

1. If  $f(n) = O(n^{c-\epsilon})$  for some  $\epsilon > 0$  =>  $\mathsf{T}(n) = \Theta(n^c)$  Intuition: the work increases as we go down the levels. Bottom level dominates the total cost.

2. If 
$$f(n) = \Theta(n^c)$$
 =>  $T(n) = \Theta(n^c \log n)$ 

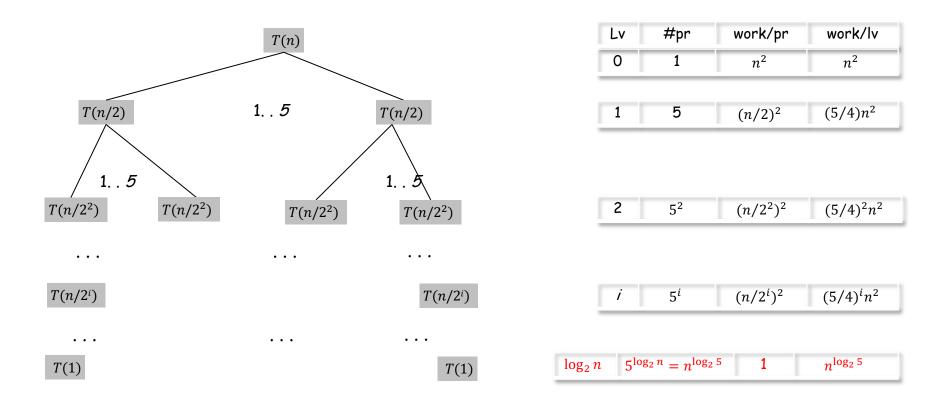
Intuition: the work remains the same as we go down the levels. All levels contribute equally to the total cost.

3. If 
$$f(n) = \Omega(n^{c+\epsilon})$$
 for some  $\epsilon > 0 \Rightarrow T(n) = \Theta(f(n))$ 

(there is an additional condition that we will ignore in this class)

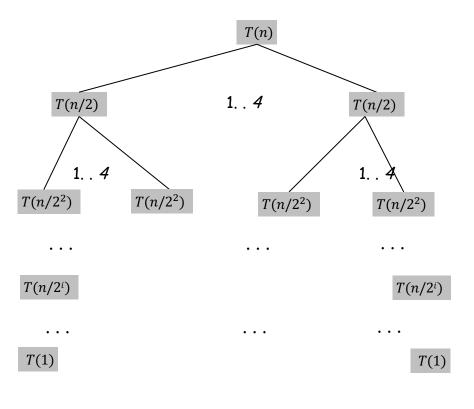
Intuition: the work decreases as we go down the levels. Top level dominates the total cost.

Case 1: 
$$T(n) = 5T(\frac{n}{2}) + n^2 = \Theta(n^{\log_2 5})$$



$$T(n) = n^{\log_2 5} + n^2 \sum_{j=0}^{\log_2 n - 1} \left(\frac{5}{4}\right)^j$$

Case 2: 
$$T(n) = 4T\left(\frac{n}{2}\right) + n^2 = \Theta(n^2 \log n)$$



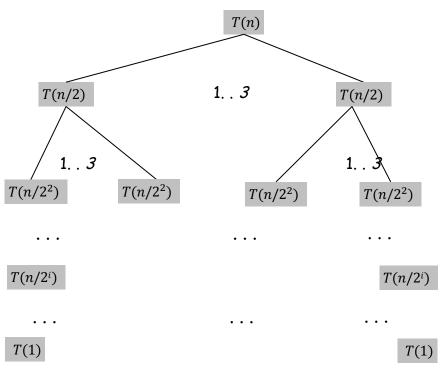
Lv	#pr	work/pr	work/lv
0	1	$n^2$	$n^2$
1	4	$(n/2)^2$	$n^2$
2	4 <sup>2</sup>	$(n/2^2)^2$	$n^2$

 $(n/2^i)^2$ 

$$\log_2 n \qquad 4^{\log_2 n} = n^{\log_2 4} \qquad \qquad 1 \qquad \qquad n^2$$
$$= n^2$$

$$T(n) = n^2 + n^2 \sum_{j=0}^{\log_2 n - 1} 1$$

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



Lv	#pr	work/pr	work/lv
0	1	$n^2$	$n^2$
1	3	$(n/2)^2$	$(3/4)n^2$

2	$3^2$	$(n/2^2)^2$	$(3/4)^2n^2$
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$$i 3^i (n/2^i)^2 (3/4)^i n^2$$

$$T(n) = n^{\log_2 3} + n^2 \sum_{j=0}^{\log_2 n - 1} \left(\frac{3}{4}\right)^j$$

## The Master Theorem (for inequalities)

$$T(n) \le aT(n/b) + f(n), \qquad c = \log_b a$$

1. If 
$$f(n) = O(n^{c-\epsilon})$$
 for some  $\epsilon > 0$  =>  $T(n) = O(n^c)$ 

2. If 
$$f(n) = O(n^c)$$
 =>  $T(n) = O(n^c \log n)$ 

3. If 
$$f(n) = \Omega(n^{c+\epsilon})$$
 for some  $\epsilon > 0$  =>  $T(n) = O(f(n))$ 

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## The Master Theorem when $f(n) = \Theta(n)$

$$T(n) = aT(n/b) + \Theta(n), \qquad c = \log_b a$$

Note: Inequality version of theorem also holds

- 1. If c > 1, then  $T(n) = \Theta(n^c)$ 
  - ightharpoonup If  $T(n) = 4T(n/2) + \Theta(n)$  then  $T(n) = \Theta(n^2)$
  - ightharpoonup If  $T(n) = 3T(n/2) + \Theta(n)$  then  $T(n) = \Theta(n^{\log_2 3}) = \Theta(n^{1.58...})$
- 2. If c = 1, then  $T(n) = \Theta(n \log n)$ 
  - ightharpoonup If  $T(n) = 2T(n/2) + \Theta(n)$  then  $T(n) = \Theta(n \log n)$
- 3. If c < 1, then  $T(n) = \Theta(n)$ 
  - ightharpoonup If  $T(n) = T(n/2) + \Theta(n)$ . then  $T(n) = \Theta(n)$

## More Master Theorem(s)

There are many variations of the Master Theorem. Here's another...

- If T(n) = T(3n/4) + T(n/5) + n then  $T(n) = \Theta(n)$
- More generally, given constants  $\alpha_i>0$  with  $\sum_i \alpha_i<1$  If  $T(n)=n+\sum_i T(\alpha_i n)$  then  $T(n)=\Theta(n)$

# The Master Theorem when f(n) = O(n)

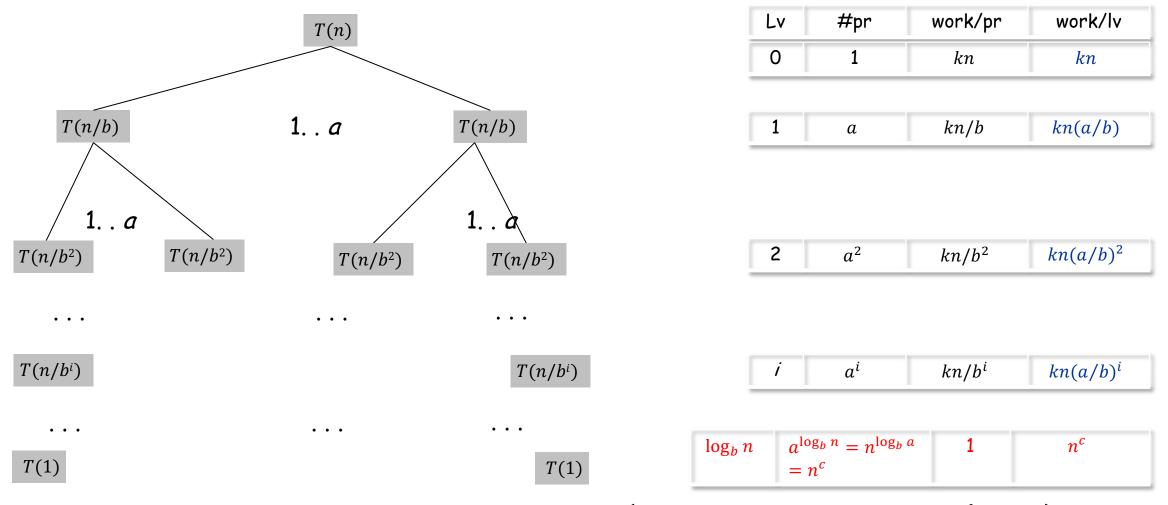
$$T(n) \le aT(n/b) + kn$$
,  $a \ge 1$  and  $b > 1$  are constants,  $c = \log_b a$ 

- 1. If c > 1, then  $T(n) = O(n^c)$
- 2. If c = 1, then  $T(n) = O(n \log n)$
- 3. If c < 1, then T(n) = O(n)

Have already worked through two examples of case 1 and one example of case 2. Will now see general proof.

# Proof of Inequality Master Theorem when f(n) = O(n)

$$T(n) \le aT(n/b) + kn$$
,  $a \ge 1$  and  $b > 1$  are constants,  $c = \log_b a$ 



• The total running time is  $n^c$  for the bottom level, plus  $kn(1+(a/b)+(a/b)^2+\cdots+(a/b)^{\log_b n-1})$  for the rest

$$c = \log_b a$$

## Case 1: a > b (c > 1)

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

If a > b, increasing geometric series

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

Hence, for a > b (c > 1)

$$T(n) = O\left(\frac{n^c}{a/b - 1}\right) = O(n^c)$$

Bottom level dominates cost.

Example: If 
$$T(n) \le 3T(\frac{n}{2}) + n$$
,  $a = 3$ ,  $b = 2$ , and  $T(n) = O(n^{\log_2 3}) = O(n^{1.58...})$ 

$$c = \log_b a$$

Case 2: 
$$a = b$$
 ( $c = 1$ )

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

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

Hence, for a = b (c = 1)

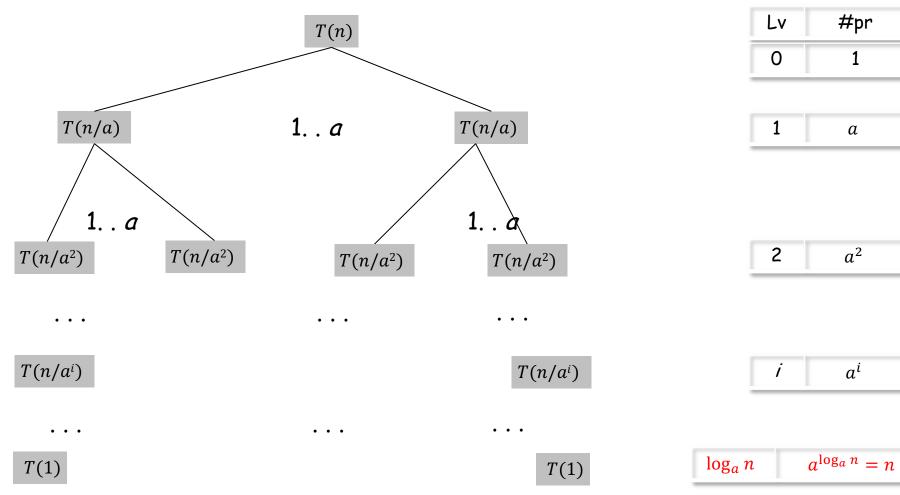
$$T(n) = O(n + kn \log_b n) = O(n \log n)$$

Each level contributes cost *kn*.

Example: If 
$$T(n) \le 2T(\frac{n}{2}) + n$$
,  $a = 2$ ,  $b = 2$ , and  $T(n) = O(n \log n)$ 

## Visualization for Case 2 (a = b)

$$T(n) \le aT(n/a) + kn, c = 1$$



0 1 kn kn   1 a kn/a kn   2 a² kn/a² kn	Lv	#pr	work/pr	work/lv
2 a <sup>2</sup> kn/a <sup>2</sup> kn	0	1	kn	kn
$2$ $a^2$ $kn/a^2$ $kn$				
,	1	а	kn/a	kn
	2	$a^2$	kn/a²	kn
$i$ $a^i$ $kn/a^i$ $kn$	í	$a^i$	kn/a <sup>i</sup>	kn
			7	

$$c = \log_b a$$

Case 3: 
$$a < b \quad (c < 1)$$

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

If a < b, decreasing geometric series

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

Hence, for a < b (c < 1)

$$T(n) = O(n^c + kn) = O(n)$$

Top level (0) dominates the cost.

Example: If 
$$T(n) \le 2T(\frac{n}{3}) + n$$
,  $a = 2$ ,  $b = 3$ , and  $T(n) = O(n)$