

Winter 2016
COMP-250: Introduction
to Computer Science

Lecture 12, February 18, 2016

Master Theorem

(CLRS 4.3)

Master Theorem

Used for many divide-and-conquer recurrences

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

where $a \geq 1$, $b > 1$, and $f(n) > 0$.

a = (constant) number of sub-instances,

b = (constant) size ration of sub-instances,

$f(n)$ = time used for dividing and recombining.

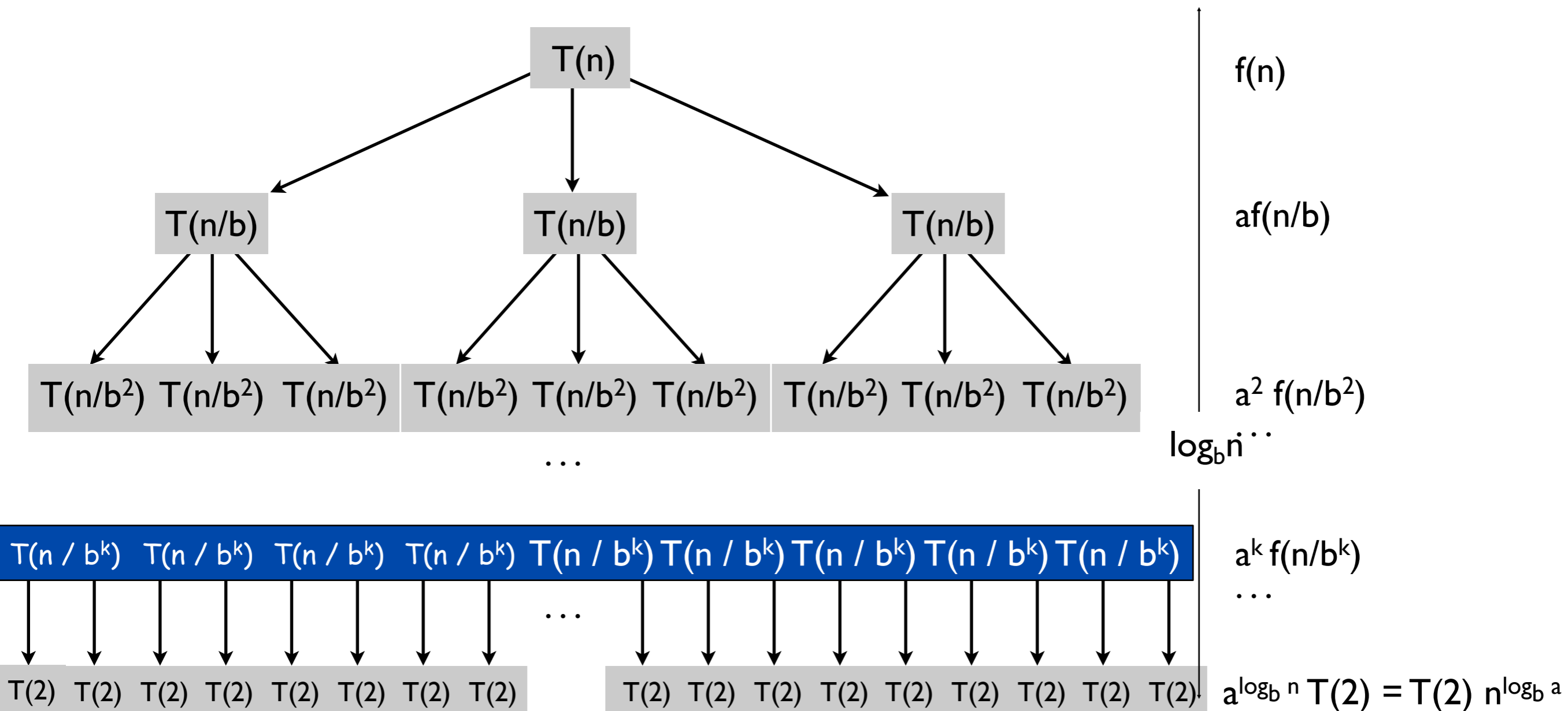
Based on the *master theorem* (Theorem 4.1).

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

Proof by recursion tree

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

$$T(n) = \sum a^k f(n/b^k)$$



Master Theorem

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

Case 1: $f(n)$ is $O(n^L)$ for some constant $L < \log_b a$.

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

Case 2: $f(n)$ is $\Theta(n^{\log_b a} \log^k n)$, for some $k \geq 0$.

Solution: $T(n)$ is $\Theta(n^{\log_b a} \log^{k+1} n)$

Case 3: $f(n)$ is $\Omega(n^L)$ for some constant $L > \log_b a$
and $f(n)$ satisfies the regularity condition $af(n/b) \leq cf(n)$ for some $c < 1$ and all large n .

Solution: $T(n)$ is $\Theta(f(n))$

Master Theorem

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

where $a \geq 1$, $b > 1$, and $f(n) > 0$.

Case 1: $f(n)$ is $O(n^L)$ for some constant $L < \log_b a$.

($f(n)$ is polynomially smaller than $n^{\log_b a}$.)

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

(Intuitively: cost is dominated by leaves.)

Master Theorem

Case 1: $f(n)$ is $O(n^L)$ for some constant $L < \log_b a$.

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

$$T(n) = 5T(n/2) + \Theta(n^2)$$

Compare $n^{\log_2 5}$ vs. n^2 .

Since $2 < \log_2 5$ use **Case 1**

Solution: $T(n)$ is $\Theta(n^{\log_2 5})$

Master Theorem

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

where $a \geq 1$, $b > 1$, and $f(n) > 0$.

Simple Case 2: $f(n)$ is $\Theta(n^{\log_b a})$.

Solution: $T(n)$ is $\Theta(n^{\log_b a} \log n)$

Master Theorem

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

where $a \geq 1$, $b > 1$, and $f(n) > 0$.

Case 2: $f(n)$ is $\Theta(n^{\log_b a} \log^k n)$, for some $k \geq 0$.

Solution: $T(n)$ is $\Theta(n^{\log_b a} \log^{k+1} n)$

(Intuitively: cost is $n^{\log_b a} \lg^k n$ at each level, and there are $\Theta(\lg n)$ levels.)

Master Theorem

Case 2: $f(n)$ is $\Theta(n^{\log_b a} \log^k n)$, for some $k \geq 0$.

Solution: $T(n)$ is $\Theta(n^{\log_b a} \log^{k+1} n)$

$$T(n) = 27T(n/3) + \Theta(n^3 \log n)$$

Compare $n^{\log_3 27}$ vs. n^3 .

Since $3 = \log_3 27$ use **Case 2**

Solution: $T(n)$ is $\Theta(n^3 \log^2 n)$

Master Theorem

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

where $a \geq 1$, $b > 1$, and $f(n) > 0$.

Case 3: $f(n)$ is $\Omega(n^L)$ for some constant $L > \log_b a$
and $f(n)$ satisfies the regularity condition $af(n/b) \leq cf(n)$ for some $c < 1$ and all large n .

($f(n)$ is polynomially greater than $n^{\log_b a}$.)

Solution: $T(n)$ is $\Theta(f(n))$

(Intuitively: cost is dominated by root.)

Master Theorem

Case 3: $f(n)$ is $\Omega(n^L)$ for some constant $L > \log_b a$
and $f(n)$ satisfies the regularity condition $af(n/b) \leq cf(n)$ for some $c < 1$ and all large n .

Solution: $T(n)$ is $\Theta(f(n))$

What's with the Case 3 regularity condition?

- Generally not a problem.
- It always holds whenever $f(n) = n^k$ and $f(n)$ is $\Omega(n^{\log_b a + \epsilon})$ for constant $\epsilon > 0$.

Master Theorem

Case 3: $f(n)$ is $\Omega(n^L)$ for some constant $L > \log_b a$
and $f(n)$ satisfies the regularity condition $af(n/b) \leq cf(n)$ for some $c < 1$ and all large n .

Solution: $T(n)$ is $\Theta(f(n))$

$$T(n) = 5T(n/2) + \Theta(n^3)$$

Compare $n^{\log_2 5}$ vs. n^3 .

Since $3 > \log_2 5$ use **Case 3**

$$af(n/b) = 5(n/2)^3 = 5/8 n^3 \leq cn^3, \text{ for } c = 5/8$$

Solution: $T(n)$ is $\Theta(n^3)$

~~Master Theorem~~

Case 2: $f(n)$ is $\Theta(n^{\log_b a} \log^k n)$, for some $k \geq 0$.

Solution: $T(n)$ is $\Theta(n^{\log_b a} \log^{k+1} n)$

$$T(n) = 27T(n/3) + \Theta(n^3/\log n)$$

Compare $n^{\log_3 27}$ vs. n^3 .

Since $3 = \log_3 27$ use **Case 2**

but $n^3/\log n$ is **not** $\Theta(n^3 \log^k n)$ for $k \geq 0$

Cannot use Master Method.

Divide-and-Conquer Paradigm

Divide-and-Conquer

Divide-and-conquer.

- Break up problem into several parts.
- Solve each part recursively.
- Combine solutions to sub-problems into overall solution.

$$T(n) = \begin{cases} 0 & \text{if } n = 1 \\ \underbrace{2T(n/2)}_{\text{sorting both halves}} + \underbrace{n}_{\text{merging}} & \text{otherwise} \end{cases}$$

Most common usage.

- Break up problem of size n into **two** equal parts of size $n/2$.
- Solve two parts recursively.
- Combine two solutions into overall solution in **linear time**.

Consequence.

- Straightforward: n^2 .
- Divide-and-conquer: $n \log n$.

Divide et impera.
Veni, vidi, vici.
- Julius Caesar

Divide-and-Conquer: Binary Search

Binary Search

Find a value v in a sorted array of elements.

$[a_0 \leq a_1 \leq \dots \leq a_{\text{size}-1}]$

Size = number of elements.

Binary Search

Algorithm: `binarySearch($a, v, low, high$)`

Input: array a , value v , lower and upper bound indices $low, high$ ($low = 0, high = n - 1$ initially)

Output: the index i of element v (if it is present), -1 (if v is not present)

```
if  $low == high$  then
  if  $a[low] == v$  then
    return  $low$ 
  else
    return  $-1$ 
  end if
else
   $mid \leftarrow (low + high) / 2$ 
  if  $v \leq a[mid]$  then
    return binarySearch( $a, v, low, mid$ )
  else
    return binarySearch( $a, v, mid + 1, high$ )
  end if
end if
```

	0	0000	L	L				
	1	0001						
	2	0010						
	3	0011						
	4	0100			L	L		
*	5	0101				H	L=H	
	6	0110						
	7	0111		H	H			
	8	1000						
	9	1001						
	10	1010						
	11	1011						
	12	1100						
	13	1101						
	14	1110						
	15	1111	H					

	0	0000	L	L			
	1	0001					
	2	0010					
	3	0011					
	4	0100			L	L	
*	5	0101				H	L=H
	6	0110					
	7	0111		H	H		
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	13	1101					
	14	1110					
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	0	0000	L	L		
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*	5	0101				H L=H
	6	0110				
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	8	1000				
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	10	1010				
	11	1011				
	12	1100				
	13	1101				
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	0	0000	L	L			
	1	0001					
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	3	0011					
	4	0100			L	L	
*	5	0101				H	L=H
	6	0110					
	7	0111		H	H		
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	12	1100					
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	14	1110					
	15	1111	H				

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	1	0001				
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*	5	0101				H L=H
	6	0110				
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	8	1000				
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	10	1010				
	11	1011				
	12	1100				
	13	1101				
	14	1110				
	15	1111	H			

Binary Search

0	L			
1				
2				
3		L	L	
* 4			H	L=H
5	H	H		

Binary Search

0	L			
1				
2				
3		L	L	
* 4			H	L=H
5	H	H		

Binary Search

0	L			
1				
2				
3		L	L	
* 4			H	L=H
5	H	H		

Binary Search

0	L			
1				
2				
3		L	L	
* 4			H	L=H
5	H	H		

Recurrence Relation

Def. $T(n)$ = number of comparisons to find v among n sorted elements.

Binary Search recurrence.

$$T(n) = \begin{cases} 1 & \text{if } n = 1 \\ T(n/2) + 1 & \text{if } n > 1 \end{cases}$$

Solution. $T(n)$ is $O(\log n)$ (Master Theorem Case 2).

Divide-and-Conquer Multiplication

Integer Multiplication

Add. Given two n -digit integers a and b , compute $a + b$.

- $\Theta(n)$ bit operations.

Multiply. Given two n -digit integers a and b , compute $a \times b$.

- Grade School solution: $\Theta(n^2)$ bit operations.

1	1	1	1	1	1	0	1	
	1	1	0	1	0	1	0	1
+	0	1	1	1	1	1	0	1
1	0	1	0	1	0	0	1	0

Add

									1	1	0	1	0	1	0	1	0		
									×	0	1	1	1	1	1	0	1		
									1	1	0	1	0	1	0	1	0		
									0	0	0	0	0	0	0	0	0		
									1	1	0	1	0	1	0	1	0		
									1	1	0	1	0	1	0	1	0		
									1	1	0	1	0	1	0	1	0		
									1	1	0	1	0	1	0	1	0		
									0	0	0	0	0	0	0	0	0		
	0	1	1	0	1	0	0	0	0	0	0	0	0	0	0	0	0	1	0

Multiply

D&C Multiplication

To multiply two n -digit integers:

- Multiply four $n/2$ -digit integers.
- Add two $n/2$ -digit integers, and shift to obtain result.

$$x = 2^{n/2} \cdot x_1 + x_0$$

$$y = 2^{n/2} \cdot y_1 + y_0$$

$$xy = (2^{n/2} \cdot x_1 + x_0)(2^{n/2} \cdot y_1 + y_0) = 2^n \cdot x_1 y_1 + 2^{n/2} \cdot (x_1 y_0 + x_0 y_1) + x_0 y_0$$

$$T(n) = \underbrace{4T(n/2)}_{\text{recursive calls}} + \underbrace{\Theta(n)}_{\text{add, shift}} \Rightarrow T(n) \text{ is } \Theta(n^2)$$



assumes n is a power of 2

Telescoping Proof

Claim.

$$T(n) = \underbrace{4T(n/2)}_{\text{recursive calls}} + \underbrace{\Theta(n)}_{\text{add, shift}} \Rightarrow T(n) \text{ is } \Theta(n^2)$$

↑
assumes n is a power of 2

Pf. For $n > 1$:

$$\begin{aligned} T(n)/n &= 4T(n/2)/n + C \\ &= 2T(n/2)/(n/2) + C \\ &= 2 [2T(n/4)/(n/4) + C] + C \\ &= 4T(n/4)/(n/4) + 2C + C \\ &= 4 [2T(n/8)/(n/8) + C] + 2C + C \\ &= 8T(n/8)/(n/8) + 4C + 2C + C \\ &\dots \\ &= nT(1)/1 + n/2 C + n/4 C + \dots + 4C + 2C + C \\ &= C (n/2 + n/4 + \dots + 2 + 1) = C(n-1). \end{aligned}$$

Karatsuba Multiplication

To multiply two n -digit integers:

- Add two $n/2$ digit integers.
- Multiply **three** $n/2$ -digit integers.
- Add, subtract, and shift $n/2$ -digit integers to obtain result.

$$\begin{aligned}x &= 2^{n/2} \cdot x_1 + x_0 \\y &= 2^{n/2} \cdot y_1 + y_0 \\xy &= 2^n \cdot x_1 y_1 + 2^{n/2} \cdot (x_1 y_0 + x_0 y_1) + x_0 y_0 \\ &= 2^n \cdot x_1 y_1 + 2^{n/2} \cdot \left((x_1 + x_0)(y_1 + y_0) - x_1 y_1 - x_0 y_0 \right) + x_0 y_0\end{aligned}$$

A B A C C

Theorem. [Karatsuba-Ofman, 1962] Can multiply two n -digit integers in $O(n^{1.585})$ bit operations.

$$T(n) \leq \underbrace{T(\lfloor n/2 \rfloor) + T(\lceil n/2 \rceil) + T(1 + \lfloor n/2 \rfloor)}_{\text{recursive calls}} + \underbrace{\Theta(n)}_{\text{add, subtract, shift}}$$

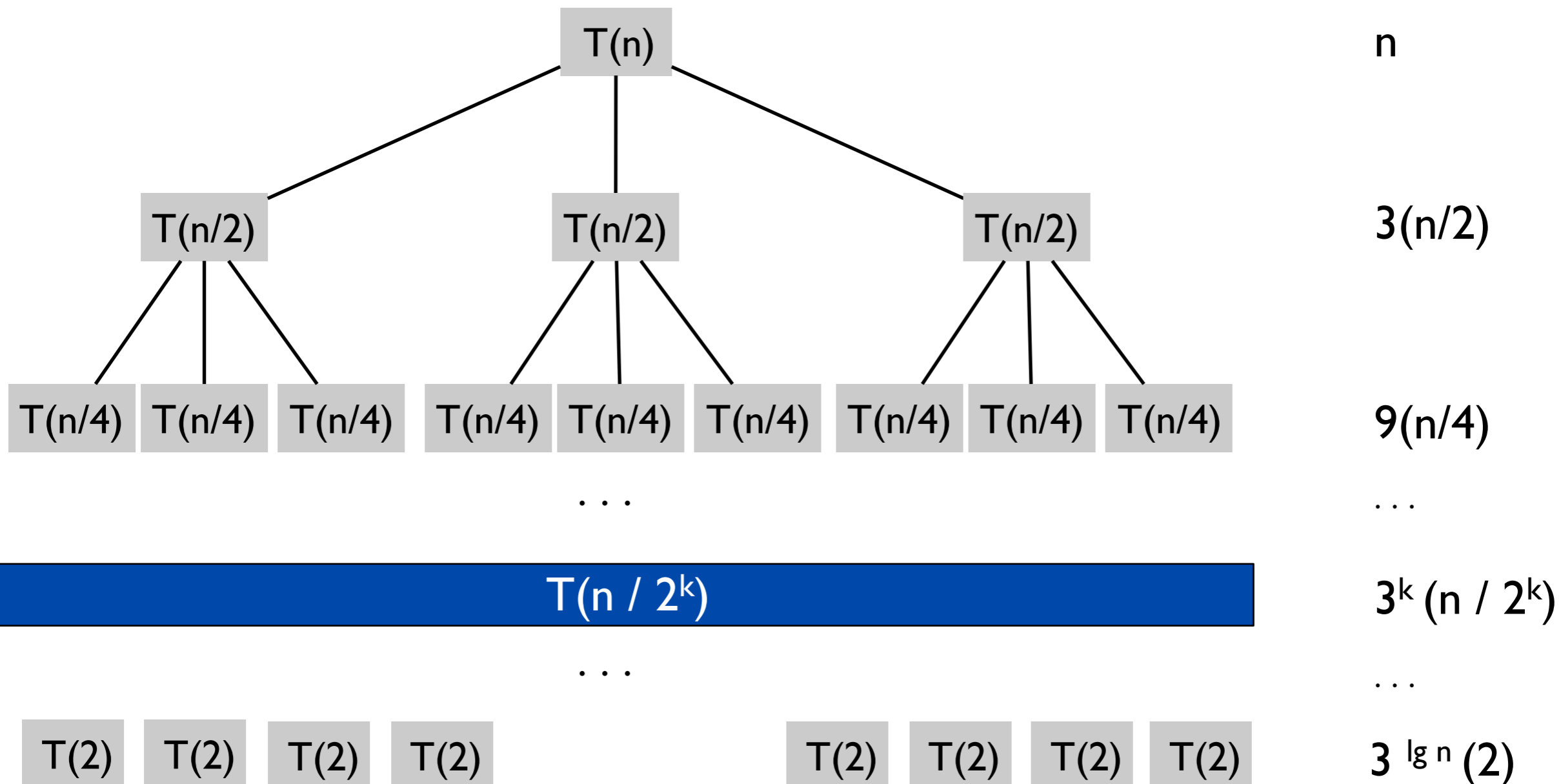
$$\Rightarrow T(n) \text{ is } O(n^{\log_2 3}) \text{ is } O(n^{1.585})$$

$$\sum_{k=0}^{n-1} ar^k = a \frac{1-r^n}{1-r}$$

Karatsuba Recursion Tree

$$T(n) = \begin{cases} 0 & \text{if } n = 1 \\ 3T(n/2) + n & \text{otherwise} \end{cases}$$

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



Karatsuba Multiplication

Generalization: $O(n^{1+\varepsilon})$ for any $\varepsilon > 0$.

Best known: $n \log n 2^{O(\log^* n)}$

where $\log^*(x) = \begin{cases} 0 & \text{if } x \leq 1 \\ 1 + \log^*(\log x) & \text{if } x > 1 \end{cases}$

Conjecture: $\Omega(n \log n)$ but not proven yet.

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