#### **Design and Analysis of Algorithm**

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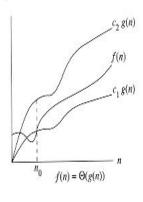
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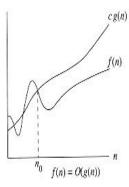


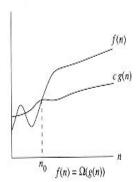
Running time is a function on domain  $N = \{0, 1, 2, \dots\}$ 

- O-notation upper bound
- Θ-notation asymptotically tight bound
- Ω-notation lower bound











#### Given two function T(n) and f(n):

- $T(n) = \mathcal{O}(f(n))$  if  $\exists$  constants c > 0 and  $n_0 \ge 0$  such that  $\forall n \ge n_0, T(n) \le c\dot{f}(n)$
- $T(n) = \Omega(f(n))$  if  $\exists$  constants c > 0 and  $n_0 \ge 0$  such that  $\forall n \ge n_0, \ T(n) \ge c\dot{f}(n)$
- $T(n) = \Theta(f(n))$  if  $T(n) = \mathcal{O}(f(n))$  and  $T(n) = \Omega(f(n))$



Let f and g be two functions such that

$$\lim_{n\to\infty}\frac{f(n)}{g(n)}=c$$

where c(>0) is a constant. Then

$$f(n) = \Theta(g(n))$$



## **Properties of Asymptotic Growth**

- 1 If  $f = \mathcal{O}(g)$  and  $g = \mathcal{O}(h)$  then  $f = \mathcal{O}(h)$
- 2 If  $f = \Omega(g)$  and  $g = \Omega(h)$  then  $f = \Omega(h)$
- 3 If  $f = \Theta(g)$  and  $g = \Theta(h)$  then  $f = \Theta(h)$
- If  $f = \mathcal{O}(h)$  and  $g = \mathcal{O}(h)$  then  $f + g = \mathcal{O}(h)$ This sum can be extended to k number of functions also
- **5** Suppose f and g are non-negative functions such that  $g = \mathcal{O}(f)$ . Then  $f + q = \Theta(f)$
- 5 For a polynomial f of degree d in which the coefficients are positive. Then  $f = \mathcal{O}(n^d)$
- Por every b > 1 and every x > 0 we have  $log_b n = \mathcal{O}(n^x)$
- 8 For every r > 1 and every d > 0 we have  $n^d = \mathcal{O}(r^n)$







A recurrence is an equation or an inequality that describes a function in terms of its value of smaller inputs.



```
MERGE-SORT (A,p,r) if p < r then q = \lfloor (p+r)/2 \rfloor MERGE-SORT (A,p,q) MERGE-SORT (A,q+1,r) MERGE (A,p,q,r)
```



$$T(n) = \begin{cases} \Theta(1) & \text{if } n \leq c, \\ aT(n/b) + D(n) + C(n) & \text{otherwise.} \end{cases}$$



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



- Involves guessing of formula
- 2 Then use the mathematical induction to show it works



Let  $\mathcal{O}(nlogn)$  be the guess for the equation T(n) = 2T(n/2) + nTo prove:  $T(n) \le cnlogn$  for a constant c > 0

$$T(n) \leq 2(cn/2log(n/2)) + n$$
  
 $\leq cnlog(n/2) + n$   
 $= cnlogn - cnlog2 + n$   
 $= cnlogn - cn + n$   
 $\leq cnlogn$ 

where c > 1



$$T(n) = 2.T(n/2) + 1$$

Let us assume the bound to be  $\mathcal{O}(n)$ . So we need to show that  $T(n) \leq cn$ 

$$T(n) \leq 2.cn/2 + 1$$
$$= cn + 1$$

$$T(n) \le (cn/2 - b) + (cn/2 - b) + 1$$
  
=  $cn - 2b + 1$   
<  $cn - b$ 



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$$T(n) \le (cn/2 - b) + (cn/2 - b) + 1$$
  
=  $cn - 2b + 1$   
<  $cn - b$ 



Guess  $T(n) \leq cn$ 

$$T(n) \le 2(cn/2) + n$$
  
 $\le cn + n$   
 $= \mathcal{O}(n)$ 

Is that correct?!



Suppose the recurrence equation is

$$T(n) = 2T(|\sqrt{n}|) + logn$$

Take m = logn

$$T(2^m) = 2T(2^{m/2}) + m$$

Take  $S(m) = T(2^m)$ 

$$S(m) = 2S(m/2) + m$$

S(m) is  $\mathcal{O}(mlogm)$ 

$$T(n) = \mathcal{O}(lognloglogn)$$



## **Recurrence Equation – Iteration Method**

Consider the iteration:

$$T(n) = 3T(n/4) + n$$

We can iterate as follows:

$$T(n) = n + 3T(n/4)$$

$$= n + 3(n/4 + 3T(n/16))$$

$$= n + 3(n/4 + 3(n/16 + 3T(n/64)))$$

$$= n + 3(n/4) + 9(n/16) + 27T(n/64),$$



## **Recurrence Equation – Iteration Method**

$$T(n) \leq n + 3n/4 + 9n/16 + 27n/64 + \dots + 3^{\log_4 n} \Theta(1)$$

$$\leq n \sum_{i=0}^{\infty} \left(\frac{3}{4}\right)^i + \Theta(n^{\log_4 3})$$

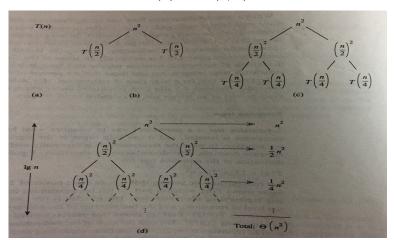
$$= 4n + o(n)$$

$$= \mathcal{O}(n).$$



### **Recurrence Equation – Recursion Trees**

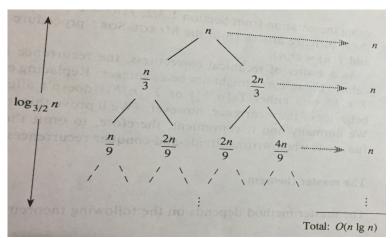
$$T(n) = 2T(n/2) + n^2$$





## **Recurrence Equation – Recursion Trees**

$$T(n) = T(n/3) + T(2n/3) + n$$





#### **Master Theorem**

Let  $a \ge 1$  and b > 1 be constants, let f(n) be a function, and let T(n) be defined on the nonnegative integers by the recurrence

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

Then T(n) can be bounded asymptotically as follows,

- If  $f(n) = \mathcal{O}(n^{log_b a \epsilon})$  for some constant  $\epsilon > 0$ , then  $T(n) = \Theta(n^{log_b a})$
- ② If  $f(n) = \Theta(n^{\log_b a})$ , then  $T(n) = \Theta(n^{\log_b a} \log n)$
- If  $f(n) = \Omega(n^{\log_b a + \epsilon})$  for some constant  $\epsilon > 0$ , and if  $af(n/b) \le cf(n)$  for some constant c < 1 and all sufficiently large n, then  $T(n) = \Theta(f(n))$



$$T(n) = 9T(n/3) + n$$
  
 $T(n) = T(2n/3) + 1$   
 $T(n) = 3T(n/4) + nlogn$ 



$$T(n) = 9T(n/3) + n$$

$$a = 9, b = 3, f(n) = n$$
 and  $n^{\log_b a} = \Theta(n^2)$ .  
Hence  $T(n) = \Theta(n^2)$ .



$$T(n) = 9T(n/3) + n$$

$$a = 9, b = 3, f(n) = n \text{ and } n^{\log_b a} = \Theta(n^2).$$
  
Hence  $T(n) = \Theta(n^2).$ 



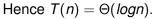
$$T(n) = T(2n/3) + 1$$

$$a = 1, b = 3/2, f(n) = 1$$
 and  $n^{\log_b a} = 1$ .  
Hence  $T(n) = \Theta(\log n)$ .



$$T(n) = T(2n/3) + 1$$

$$a = 1, b = 3/2, f(n) = 1$$
 and  $n^{\log_b a} = 1$ .





$$T(n) = 3T(n/4) + nlogn$$

a=3, b=4, f(n)=nlogn and  $n^{log_ba}=\mathcal{O}(n^{0.793})$ . Also, For larger n,  $af(n/b)=3(n/4)log(n/4)\leq (3/4)nlogn=cf(n)$  for c=3/4. Hence  $T(n)=\Theta(nlogn)$ .



$$T(n)=3T(n/4)+nlogn$$
  $a=3, b=4, f(n)=nlogn$  and  $n^{\log_b a}=\mathcal{O}(n^{0.793}).$  Also, For larger  $n$ ,  $af(n/b)=3(n/4)log(n/4)\leq (3/4)nlogn=cf(n)$  for  $c=3/4.$  Hence  $T(n)=\Theta(nlogn).$ 



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

$$a = 2, b = 2, f(n) = nlogn \text{ and } n^{\log_b a} = n.$$

But *nlogn* is asymptotically larger than *n* but not polynomially! So Master theorem is not applicable to this recurrence relation.



$$T(n) = aT(n/b) + n^c$$

where  $a, b \ge 1$  and c > 0 then

$$T(n) = egin{cases} \Theta(n^{log_b a}) & \text{if } a > b^c, \ \Theta(n^c log_b n) & \text{if } a = b^c, \ \Theta(n^c) & \text{if } a < b^c. \end{cases}$$



$$T(n) = aT(\frac{n}{b}) + n^{c}$$

$$= n^{c} + a((\frac{n}{b})^{c} + aT(\frac{n}{b^{2}}))$$

$$= n^{c} + (\frac{a}{b^{c}})n^{c} + a^{2}T(\frac{n}{b^{2}})$$

$$= \cdots$$

$$= n^{c} + (\frac{a}{b^{c}})n^{c} + (\frac{a}{b^{c}}))^{2}n^{c} + (\frac{a}{b^{c}}))^{3}n^{c} + \cdots$$

$$(\frac{a}{b^{c}}))^{log_{b}n-1}n^{c} + a^{log_{b}n}T(1)$$



#### **Case 1:** $a < b^c$

$$a < b^{c} \iff \frac{a}{b^{c}} < 1 \implies \sum_{k=0}^{\log_{b} n-1} \left(\frac{a}{b^{c}}\right)^{k}$$

$$\leq \sum_{k=0}^{\infty} \left(\frac{a}{b^{c}}\right)^{k} = \frac{1}{1 - \left(\frac{a}{b^{c}}\right)} = \Theta(1)$$

$$a < b^{c} \iff \log_{b} a < \log_{b} b^{c} = c$$

$$T(n) = n^{c} \sum_{k=0}^{\log_{b} n-1} \left(\frac{a}{b^{c}}\right)^{k} + n^{\log_{b} a}$$

$$= n^{c} \cdot \Theta(1) + n^{\log_{b} a}$$

$$= \Theta(n^{c})$$



**Case 2:**  $a = b^{c}$ 

$$a = b^{c} \iff \frac{a}{b^{c}} = 1 \implies \sum_{k=0}^{\log_{b} n - 1} \left(\frac{a}{b^{c}}\right)^{k}$$

$$= \sum_{k=0}^{\log_{b} n - 1} 1 = \Theta(\log_{b} n)$$

$$a = b^{c} \iff \log_{b} a = \log_{b} b^{c} = c$$

$$T(n) = n^{c} \sum_{k=0}^{\log_{b} n - 1} \left(\frac{a}{b^{c}}\right)^{k} + n^{\log_{b} a}$$

$$= n^{c} \cdot \Theta(\log_{b} n) + n^{\log_{b} a}$$

$$= \Theta(n^{c} \log_{b} n)$$





#### **Case 3:** $a > b^{c}$

$$a > b^{c} \iff \frac{a}{b^{c}} > 1 \implies \sum_{k=0}^{log_{b}n-1} (\frac{a}{b^{c}})^{k}$$

$$= \Theta((\frac{a}{b^{c}})^{log_{b}n}) = \Theta(\frac{a^{log_{b}n}}{(b^{c})^{log_{b}n}}) = \Theta(\frac{a^{log_{b}n}}{n^{c}})$$

$$T(n) = n^{c} \cdot \Theta(\frac{a^{log_{b}n}}{n^{c}}) + n^{log_{b}a}$$

$$= \Theta(n^{log_{b}a}) + n^{log_{b}a}$$

$$= \Theta(n^{log_{b}a})$$

