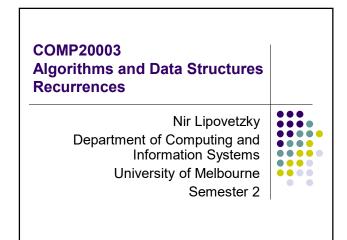
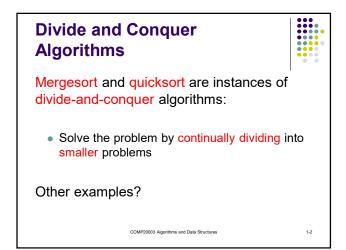
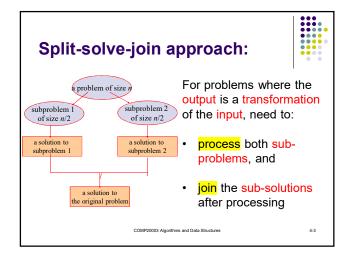
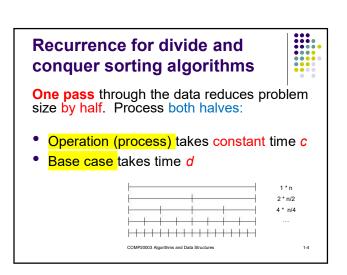
# Recurrences and Master Theorem









### Recurrence for divide and conquer sorting algorithms



One pass through the data reduces problem size by half. Process both halves

- Operation takes constant time c
- Base case takes time d

$$T(1) = d$$
  
 $T(n) = 2T(n/2) + nc$   
 $= nc + 2cn/2 + 4cn/4... + n/2*2c + nd$   
 $= c(n-1)log n + nd$ 

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### **Divide and Conquer: Recurrences to Master Theorem**



Most common case:

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

General case:

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

 $f(n) \in \Theta(n^{\mathbf{d}})$ 

Most common case:

$$T(n) = 2T(n/2) + n$$
  
 $a=2, b=2, d=1$ 

### **Master Theorem for Divide and** Conquer



- T(n) = aT(n/b) + f(n) $f(n) \in \Theta(n^d)$
- *T(n)* closed form varies, depending on whether:
  - $T(n) \in \Theta(n^d)$  $d > \log_{b}a$
  - $d = log_b a$  $T(n) \in \Theta(n^d \log n)$
  - $T(n) \in \Theta(n^{\log_b a})$  $d < log_b a$

### **Master Theorem for Divide and** Conquer



- T(n) = aT(n/b) + f(n), where a>1, b>1, nd asymptotically positive
- *T(n)* closed form varies, depending on whether:
  - $T(n) \in \Theta(n^d)$  $d > log_b a$
  - $d = log_b a$  $T(n) \in \Theta(n^d \log n)$
  - $d < log_b a$  $T(n) \in \Theta(n^{\log_b a})$

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# Recurrences and Master Theorem

### Where do $\Theta$ () solutions to the **Master Theorem come from?**



 $T(n) = aT(n/b) + f(n), f(n) \in \Theta(n^d)$ 

Size of subproblems decreases by b

- So base case reached after log<sub>b</sub>n levels
- Recursion tree log<sub>b</sub>n levels

### Branch factor is a

At kth level, have ak subproblems

At level k, total work is then

- $a^k * O(n/b^k)^d$
- (#subproblems \* cost of solving one)

### Where do ⊕() solutions to the **Master Theorem come from?**



 $T(n) = aT(n/b) + f(n), f(n) \in \Theta(n^d)$ 

- •At level k, total work is then
  - $a^k * O(n/b^k)^d = O(n^d) * (a/b^d)^k$
- As k (levels) goes from 0 to log<sub>b</sub>n, this is a geometric series, with ratio a/bd
  - $\Sigma O(n^d)^* (a/b^d)^k$

### Where do $\Theta$ () solutions to the **Master Theorem come from?**



 $T(n) = aT(n/b) + f(n), f(n) \in \Theta(n^d)$ 

- Geometric series: O(n<sup>d</sup>) \* (a/b<sup>d</sup>)<sup>k</sup>
  - as k goes from 0 → log<sub>b</sub>n
- •Case 1: ratio a/b<sup>d</sup>< 1
  - $(a/b^d)^k$  gets smaller as k goes from 1  $\rightarrow$  log n
  - a/bd First term is the largest, and is <1
  - O(n<sup>d</sup>)

## Example for $a/b^d < 1$



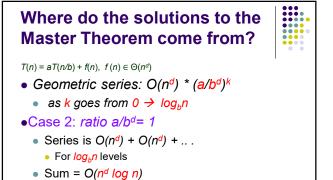
 $T(n) = 2T(n/2) + n^2$ 



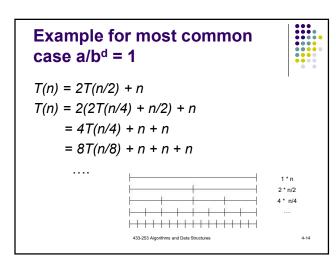
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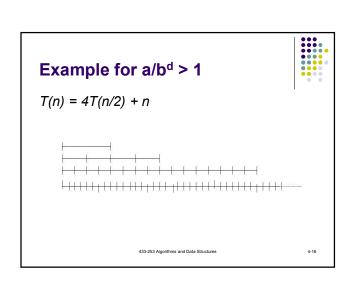
# Recurrences and Master Theorem



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# Where do $\Theta$ () solutions to the Master Theorem come from? $T(n) = aT(n/b) + f(n), \ f(n) \in \Theta(n^d)$ • Geometric series: $O(n^d) * (a/b^d)^k$ • as k goes from $0 \to log_b n$ • Case 3: ratio $a/b^d > 1$ • $a/b^d > 1 \to series$ is increasing • Sum dominated by last term: • $O(n^d)(a/b^d)^{log(b)n} = n^{log(b)a}$



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# Recurrences and Master Theorem



• For more on geometric series, and calculation of closed form, see:

http://www.youtube.com/watch?v=JJZ-shHiayU

• 4 minute tutorial from Rose-Hulman Institute of Technology

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