

CSE 417: Algorithms and Computational Complexity

Winter 2006

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Lectures 16-19

Divide and Conquer Algorithms

1

The Divide and Conquer Paradigm

- Outline:
 - General Idea
 - Review of Merge Sort
 - Why does it work?
 - Importance of balance
 - Importance of super-linear growth
 - Two interesting applications
 - Polynomial Multiplication
 - Matrix Multiplication
 - Finding & Solving Recurrences

2

Algorithm Design Techniques

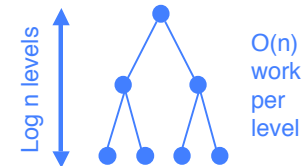
- Divide & Conquer
 - Reduce problem to one or more sub-problems of the same type
 - Typically, each sub-problem is at most a constant fraction of the size of the original problem
 - e.g. Mergesort, Binary Search, Strassen's Algorithm, Quicksort (kind of)

3

Mergesort (review)

Mergesort: (recursively) sort 2 half-lists, then merge results.

- $T(n) = 2T(n/2) + cn, n \geq 2$
- $T(1) = 0$
- Solution: $\Theta(n \log n)$ (details later)



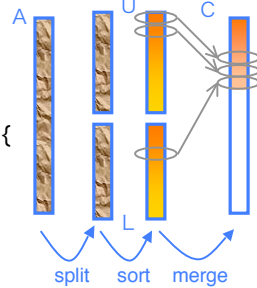
4

Merge Sort

```

MS(A: array[1..n]) returns array[1..n] {
  If(n=1) return A[1];
  New U:array[1:n/2] = MS(A[1..n/2]);
  New L:array[1:n/2] = MS(A[n/2+1..n]);
  Return(Merge(U,L));
}

Merge(U,L: array[1..n]) {
  New C: array[1..2n];
  a=1; b=1;
  For i = 1 to 2n
    C[i] = "smaller of U[a], L[b] and correspondingly a++ or b++";
  Return C;
}
    
```



5

Going From Code to Recurrence

- Carefully define what you're counting, and write it down!
 "Let $C(n)$ be the number of comparisons between sort keys used by MergeSort when sorting a list of length $n \geq 1$ "
- In code, clearly separate **base case** from **recursive case**, highlight **recursive calls**, and **operations being counted**.
- Write Recurrence(s)

6

Merge Sort

```

MS(A: array[1..n]) returns array[1..n] {
  If(n=1) return A[1];
  New L:array[1:n/2] = MS(A[1..n/2]);
  New R:array[1:n/2] = MS(A[n/2+1..n]);
  Return(Merge(L,R));
}

Merge(A,B: array[1..n]) {
  New C: array[1..2n];
  a=1; b=1;
  For i = 1 to 2n {
    C[i] = "smaller of A[a], B[b] and a++ or b++";
  }
  Return C;
}
    
```

Base Case

Recursive calls

Recursive case

Operations being counted

7

The Recurrence

$$C(n) = \begin{cases} 0 & \text{if } n = 1 \\ 2C(n/2) + (n - 1) & \text{if } n > 1 \end{cases}$$

Base case

Recursive calls

One compare per element added to merged list, except the last.

Total time: proportional to $C(n)$

(loops, copying data, parameter passing, etc.)

8

Why Balanced Subdivision?

- Alternative "divide & conquer" algorithm:
 - Sort n-1
 - Sort last 1
 - Merge them
- $T(n)=T(n-1)+T(1)+3n$ for $n \geq 2$
- $T(1)=0$
- Solution: $3n + 3(n-1) + 3(n-2) \dots = \Theta(n^2)$

9

Another D&C Approach

- Suppose we've already invented DumbSort, taking time n^2
- Try *Just One Level* of divide & conquer:
 - DumbSort(first $n/2$ elements)
 - DumbSort(last $n/2$ elements)
 - Merge results
- Time: $2 (n/2)^2 + n = n^2/2 + n \ll n^2$ D&C in a nutshell
 - Almost twice as fast!

10

Another D&C Approach, cont.

- Moral 1: "two halves are better than a whole"
Two problems of half size are *better* than one full-size problem, even given the $O(n)$ overhead of recombining, since the base algorithm has *super-linear* complexity.
- Moral 2: "If a little's good, then more's better"
two levels of D&C would be almost 4 times faster, 3 levels almost 8, etc., even though overhead is growing. Best is usually full recursion down to some small constant size (balancing "work" vs "overhead").

11

Another D&C Approach, cont.

- Moral 3: unbalanced division less good:
 - $(.1n)^2 + (.9n)^2 + n = .82n^2 + n$
 - The 18% savings compounds significantly if you carry recursion to more levels, actually giving $O(n \log n)$, but with a bigger constant. So worth doing if you can't get 50-50 split, but balanced is better if you can.
 - This is intuitively why Quicksort with random splitter is good – badly unbalanced splits are rare, and not instantly fatal.
 - $(1)^2 + (n-1)^2 + n = n^2 - 2n + 2 + n$
 - Little improvement here.

12

5.4 Closest Pair of Points

Closest Pair of Points

Closest pair. Given n points in the plane, find a pair with smallest Euclidean distance between them.

Fundamental geometric primitive.

- Graphics, computer vision, geographic information systems, molecular modeling, air traffic control.
 - Special case of nearest neighbor, Euclidean MST, Voronoi.
- ↑
fast closest pair inspired fast algorithms for these problems

Brute force. Check all pairs of points p and q with $\Theta(n^2)$ comparisons.

1-D version. $O(n \log n)$ easy if points are on a line.

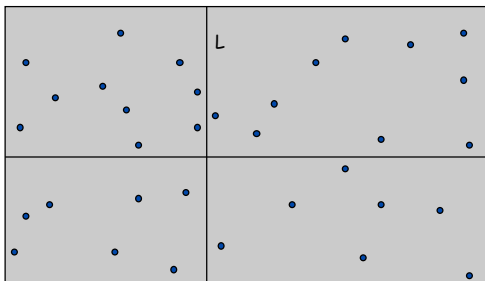
Assumption. No two points have same x coordinate.

↑
to make presentation cleaner

14

Closest Pair of Points: First Attempt

Divide. Sub-divide region into 4 quadrants.

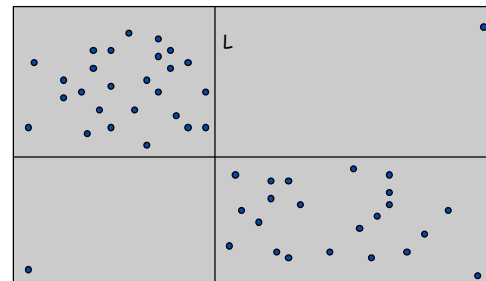


15

Closest Pair of Points: First Attempt

Divide. Sub-divide region into 4 quadrants.

Obstacle. Impossible to ensure $n/4$ points in each piece.

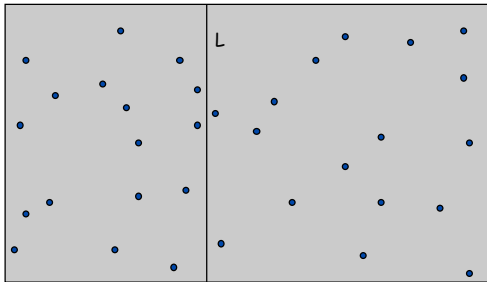


16

Closest Pair of Points

Algorithm.

- **Divide:** draw vertical line L so that roughly $\frac{1}{2}n$ points on each side.

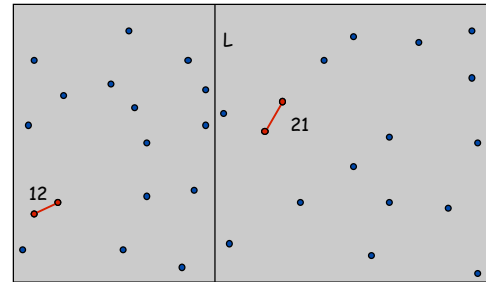


17

Closest Pair of Points

Algorithm.

- **Divide:** draw vertical line L so that roughly $\frac{1}{2}n$ points on each side.
- **Conquer:** find closest pair in each side recursively.

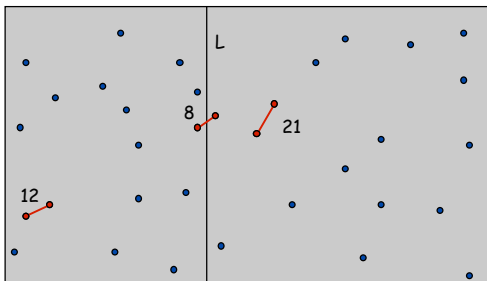


18

Closest Pair of Points

Algorithm.

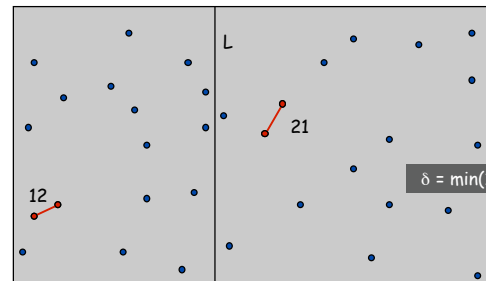
- **Divide:** draw vertical line L so that roughly $\frac{1}{2}n$ points on each side.
- **Conquer:** find closest pair in each side recursively.
- **Combine:** find closest pair with one point in each side. ← seems like $\Theta(n^2)$
- Return best of 3 solutions.



19

Closest Pair of Points

Find closest pair with one point in each side, assuming that distance $< \delta$.

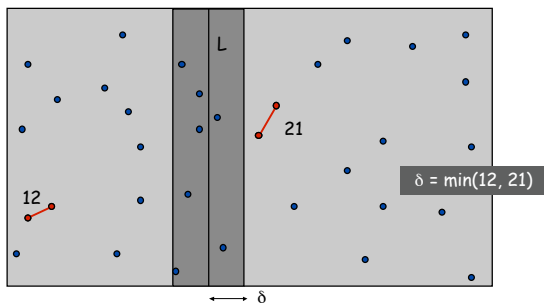


20

Closest Pair of Points

Find closest pair with one point in each side, assuming that distance $< \delta$.

- Observation: only need to consider points within δ of line L .

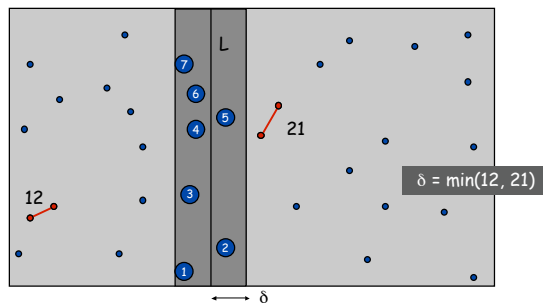


21

Closest Pair of Points

Find closest pair with one point in each side, assuming that distance $< \delta$.

- Observation: only need to consider points within δ of line L .
- Sort points in 2δ -strip by their y coordinate.

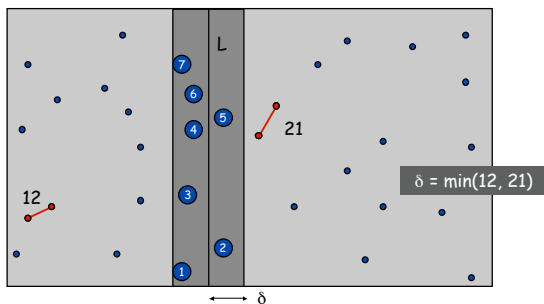


22

Closest Pair of Points

Find closest pair with one point in each side, assuming that distance $< \delta$.

- Observation: only need to consider points within δ of line L .
- Sort points in 2δ -strip by their y coordinate.
- Only check distances of those within 11 positions in sorted list!



23

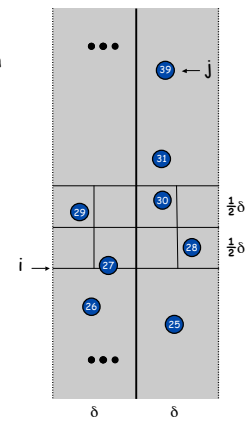
Closest Pair of Points

Def. Let s_i be the point in the 2δ -strip, with the i^{th} smallest y -coordinate.

Claim. If $|i - j| \geq 8$, then the distance between s_i and s_j is at least δ .

Pf.

- No two points lie in same $\frac{1}{2}\delta$ -by- $\frac{1}{2}\delta$ box.
- only 8 boxes



24

Closest Pair Algorithm

```

Closest-Pair( $p_1, \dots, p_n$ ) {
  if ( $n \leq ??$ ) return ??

  Compute separation line L such that half the points
  are on one side and half on the other side.

   $\delta_1 = \text{Closest-Pair}(\text{left half})$ 
   $\delta_2 = \text{Closest-Pair}(\text{right half})$ 
   $\delta = \min(\delta_1, \delta_2)$ 

  Delete all points further than  $\delta$  from separation line L

  Sort remaining points  $p[1]..p[m]$  by y-coordinate.

  for  $i = 1..m$ 
     $k = 1$ 
    while  $i+k \leq m$  &&  $p[i+k].y < p[i].y + \delta$ 
       $\delta = \min(\delta, \text{distance between } p[i] \text{ and } p[i+k]);$ 
       $k++$ ;

  return  $\delta$ .
}

```

25

Going From Code to Recurrence

- Carefully define what you're counting, and write it down!
 "Let $C(n)$ be the number of comparisons between sort keys used by MergeSort when sorting a list of length $n \geq 1$ "
- In code, clearly separate **base case** from **recursive case**, highlight **recursive calls**, and **operations being counted**.
- Write Recurrence(s)

26

Closest Pair Algorithm

Base Case

Recursive calls (2)

Basic operations: distance calcs

```

Closest-Pair( $p_1, \dots, p_n$ ) {
  if ( $n \leq 1$ ) return  $\infty$ 

  Compute separation line L such that half the points
  are on one side and half on the other side.

   $\delta_1 = \text{Closest-Pair}(\text{left half})$ 
   $\delta_2 = \text{Closest-Pair}(\text{right half})$ 
   $\delta = \min(\delta_1, \delta_2)$ 

  Delete all points further than  $\delta$  from separation line L

  Sort remaining points  $p[1]..p[m]$ 

  Basic operations at this recursive level

  for  $i = 1..m$ 
     $k = 1$ 
    while  $i+k \leq m$  &&  $p[i+k].y < p[i].y + \delta$ 
       $\delta = \min(\delta, \text{distance between } p[i] \text{ and } p[i+k]);$ 
       $k++$ ;

  return  $\delta$ .
}

```

0

$2T(n/2)$

$O(n)$

27

Closest Pair of Points: Analysis

Running time.

$$T(n) \leq \begin{cases} 0 & n = 1 \\ 2T(n/2) + O(n) & n > 1 \end{cases} \Rightarrow T(n) = O(n \log n)$$

BUT - that's only the number of distance calculations

28

Closest Pair Algorithm

Base Case

Basic operations: comparisons

```

Closest-Pair( $p_1, \dots, p_n$ ) {
  if ( $n \leq 1$ ) return  $\infty$ 
  Recursive calls (2)
  Compute separation line  $L$  such that half the points
  are on one side and half on the other side.
   $\delta_1 = \text{Closest-Pair}(\text{left half})$ 
   $\delta_2 = \text{Closest-Pair}(\text{right half})$ 
   $\delta = \min(\delta_1, \delta_2)$ 
  Delete all points further than  $\delta$  from separation line  $L$ 
  Sort remaining points  $p[1] \dots p[m]$ 
  Basic operations at this recursive level
  for  $i = 1 \dots m$ 
     $k = 1$ 
    while  $i+k \leq m$  &&  $p[i+k].y < p[i].y + \delta$ 
       $\delta = \min(\delta, \text{distance between } p[i] \text{ and } p[i+k]);$ 
       $k++$ ;
  return  $\delta$ .
}
  
```

0
 $O(n \log n)$
 $2T(n/2)$
1
 $O(n)$
 $O(n \log n)$
 $O(n)$

29

Closest Pair of Points: Analysis

Running time.

$$T(n) \leq \begin{cases} 0 & n=1 \\ 2T(n/2) + O(n \log n) & n > 1 \end{cases} \Rightarrow T(n) = O(n \log^2 n)$$

Q. Can we achieve $O(n \log n)$?

A. Yes. Don't sort points from scratch each time.

- Sort by x at top level only.
- Each recursive call returns δ and list of all points sorted by y
- Sort by **merging** two pre-sorted lists.

$$T(n) \leq 2T(n/2) + O(n) \Rightarrow T(n) = O(n \log n)$$

30

5.5 Integer Multiplication

Integer Arithmetic

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

- $O(n)$ bit operations.

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

- Brute force solution: $\Theta(n^2)$ bit operations.

```

1 1 1 1 1 1 0 1
+ 0 1 1 1 1 1 0 1
-----
1 0 1 0 1 0 0 1 0
Add
      
```

Multiply

```

      1 1 0 1 0 1 0 1
      * 0 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 1 0
      
```

32

Divide-and-Conquer Multiplication: Warmup

To multiply two n -digit integers:

- Multiply four $\frac{1}{2}n$ -digit integers.
- Add two $\frac{1}{2}n$ -digit integers, and shift to obtain result.

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

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

↑
assumes n is a power of 2

33

Karatsuba Multiplication

To multiply two n -digit integers:

- Add two $\frac{1}{2}n$ digit integers.
- Multiply **three** $\frac{1}{2}n$ -digit integers.
- Add, subtract, and shift $\frac{1}{2}n$ -digit integers to obtain result.

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

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

$$\begin{aligned} T(n) &\leq \underbrace{T(\lfloor n/2 \rfloor)}_{\text{recursive calls}} + \underbrace{T(\lceil n/2 \rceil)}_{\text{recursive calls}} + \underbrace{\Theta(n)}_{\text{add, subtract, shift}} \\ \text{Sloppy version: } T(n) &\leq 3T(n/2) + O(n) \\ \Rightarrow T(n) &= O(n^{\log_2 3}) = O(n^{1.585}) \end{aligned}$$

34

Multiplication – The Bottom Line

- Naïve: $\Theta(n^2)$
- Karatsuba: $\Theta(n^{1.59\dots})$
- Amusing exercise: generalize Karatsuba to do 5 size $n/3$ subproblems $\Rightarrow \Theta(n^{1.46\dots})$
- Best known: $\Theta(n \log n \log \log n)$
 - "Fast Fourier Transform"
 - but mostly unused in practice (unless you need really big numbers)

35

Recurrences

- Where they come from, how to find them (above)
- Next: how to solve them

36

Mergesort (review)

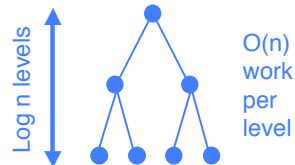
Mergesort: (recursively) sort 2 half-lists, then merge results.

■ $T(n) = 2T(n/2) + cn, n \geq 2$

■ $T(1) = 0$

■ Solution: $\Theta(n \log n)$
(details later)

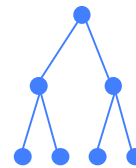
now



37

Solve: $T(1) = c$

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



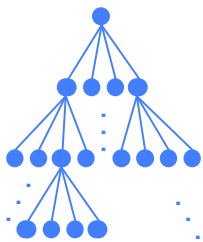
Level	Num	Size	Work
0	$1=2^0$	n	cn
1	$2=2^1$	$n/2$	$2 c n/2$
2	$4=2^2$	$n/4$	$4 c n/4$
...
i	2^i	$n/2^i$	$2^i c n/2^i$
...
$k-1$	2^{k-1}	$n/2^{k-1}$	$2^{k-1} c n/2^{k-1}$
k	2^k	$n/2^k=1$	$2^k T(1)$

Total work: add last col

38

Solve: $T(1) = c$

$$T(n) = 4 T(n/2) + cn$$



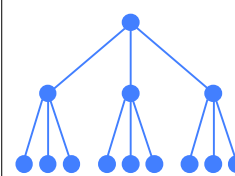
Level	Num	Size	Work
0	$1=4^0$	n	cn
1	$4=4^1$	$n/2$	$4 c n/2$
2	$16=4^2$	$n/4$	$16 c n/4$
...
i	4^i	$n/2^i$	$4^i c n/2^i$
...
$k-1$	4^{k-1}	$n/2^{k-1}$	$4^{k-1} c n/2^{k-1}$
k	4^k	$n/2^k=1$	$4^k T(1)$

$$\sum_{i=0}^k 4^i cn / 2^i = O(n^2)$$

39

Solve: $T(1) = c$

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



Level	Num	Size	Work
0	$1=3^0$	n	cn
1	$3=3^1$	$n/2$	$3 c n/2$
2	$9=3^2$	$n/4$	$9 c n/4$
...
i	3^i	$n/2^i$	$3^i c n/2^i$
...
$k-1$	3^{k-1}	$n/2^{k-1}$	$3^{k-1} c n/2^{k-1}$
k	3^k	$n/2^k=1$	$3^k T(1)$

$n = 2^k ; k = \log_2 n$

$$\text{Total Work: } T(n) = \sum_{i=0}^k 3^i cn / 2^i$$

40

Solve: $T(1) = c$
 $T(n) = 3 T(n/2) + cn$ (cont.)

$$\begin{aligned}
 T(n) &= \sum_{i=0}^k 3^i cn / 2^i \\
 &= cn \sum_{i=0}^k 3^i / 2^i \\
 &= cn \sum_{i=0}^k \left(\frac{3}{2}\right)^i \\
 &= cn \frac{\left(\frac{3}{2}\right)^{k+1} - 1}{\left(\frac{3}{2}\right) - 1}
 \end{aligned}$$

$$\begin{aligned}
 \sum_{i=0}^k x^i &= \\
 \frac{x^{k+1} - 1}{x - 1} \\
 (x \neq 1)
 \end{aligned}$$

41

Solve: $T(1) = c$
 $T(n) = 3 T(n/2) + cn$ (cont.)

$$\begin{aligned}
 &= 2cn \left(\left(\frac{3}{2}\right)^{k+1} - 1 \right) \\
 &< 2cn \left(\frac{3}{2}\right)^{k+1} \\
 &= 3cn \left(\frac{3}{2}\right)^k \\
 &= 3cn \frac{3^k}{2^k}
 \end{aligned}$$

42

Solve: $T(1) = c$
 $T(n) = 3 T(n/2) + cn$ (cont.)

$$\begin{aligned}
 &= 3cn \frac{3^{\log_2 n}}{2^{\log_2 n}} \\
 &= 3cn \frac{3^{\log_2 n}}{n} \\
 &= 3c 3^{\log_2 n} \\
 &= 3c (n^{\log_2 3}) \\
 &= O(n^{1.59...})
 \end{aligned}$$

$$\begin{aligned}
 a^{\log_b n} &= \\
 &= (b^{\log_b a})^{\log_b n} \\
 &= (b^{\log_b n})^{\log_b a} \\
 &= n^{\log_b a}
 \end{aligned}$$

43

Master Divide and Conquer Recurrence

- If $T(n) = aT(n/b) + cn^k$ for $n > b$ then
 - if $a > b^k$ then $T(n)$ is $\Theta(n^{\log_b a})$ [many subproblems => leaves dominate]
 - if $a < b^k$ then $T(n)$ is $\Theta(n^k)$ [few subproblems => top level dominates]
 - if $a = b^k$ then $T(n)$ is $\Theta(n^k \log n)$ [balanced => all log n levels contribute]
- Works even if it is $\lceil n/b \rceil$ instead of n/b .

44

D & C Summary

- “two halves are better than a whole”
if the base algorithm has super-linear complexity.
- “If a little's good, then more's better”
repeat above, recursively
- Analysis: recursion tree or Master Recurrence

45