

**CSE 421**

# **Introduction to Algorithms**

## **Lecture 8: Divide and Conquer**

# Algorithm Design Techniques

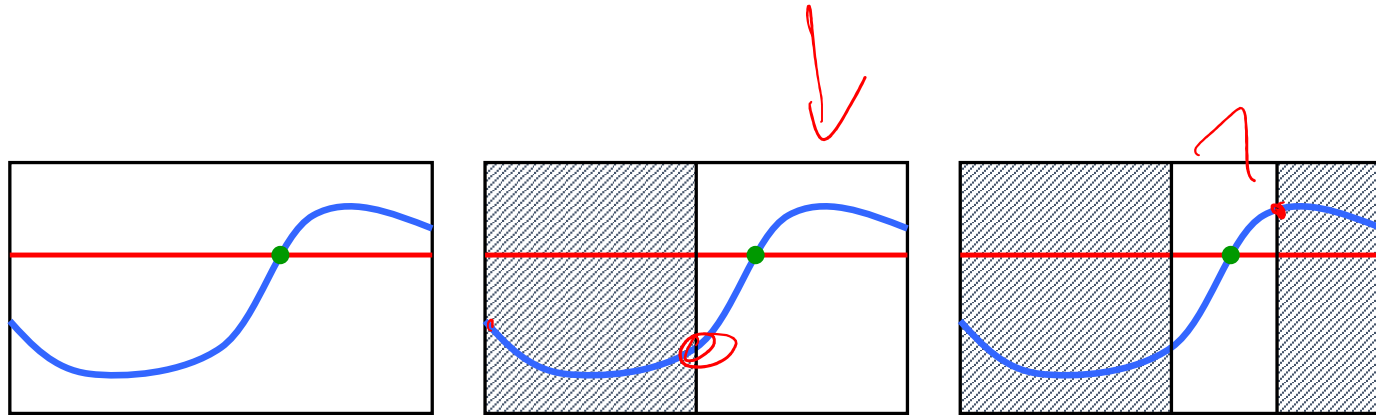
## Divide & Conquer

- Divide instance into subparts.
- Solve the parts recursively.
- Conquer by combining the answers

To truly fit Divide & Conquer

- each sub-part should be **at most a constant fraction** of the size of the original input instance
  - e.g. Mergesort, Binary Search, Quicksort (sort of), etc.

## Binary search for roots (bisection method)



### Given:

- Continuous function  $f$  and two points  $a < b$  with  $f(a) \leq 0$  and  $f(b) > 0$

### Find:

- Approximation within  $\epsilon$  of  $c$  s.t.  $f(c) = 0$  and  $a < c < b$

# Bisection method

Bisection( $a$ ,  $b$ ,  $\epsilon$ )

if  $(b - a) \leq \epsilon$  then

    return( $a$ )

else {

$c \leftarrow (a + b)/2$

    if  $f(c) \leq 0$  then

        return(Bisection( $c$ ,  $b$ ,  $\epsilon$ ))

    else

        return(Bisection( $a$ ,  $c$ ,  $\epsilon$ ))

}

# Time Analysis

At each step we halved the size of the interval

- It started at size  $b - a$
- It ended at size  $\epsilon$

So # of calls to  $f$  is  $\log_2 ((b - a)/\epsilon)$

# Old Favorites

## Binary search:

- One subproblem of half size plus one comparison
- Recurrence\* for time in terms of # of comparisons
- $T(n) = T(n/2) + 1$  for  $n \geq 2$
- $T(1) = 0$
- Solving shows that  $T(n) = \lceil \log_2 n \rceil + 1$

## Mergesort:

- Two subproblems of half size plus merge cost of  $n - 1$  comparisons
- Recurrence\* for time in terms of # of comparisons
  - $T(n) \leq 2T(n/2) + n - 1$  for  $n \geq 2$
  - $T(1) = 0$
- Roughly  $n$  comparisons at each of  $\log_2 n$  levels of recursion so  $T(n)$  is roughly  $n \log_2 n$

\*We will implicitly assume that every input to  $T(\cdot)$  is rounded up to the nearest integer.

# Euclidean Closest Pair

## Given:

- A sequence of  $n$  points  $p_1, \dots, p_n$  with real coordinates in  $d$  dimensions ( $\mathbb{R}^d$ )

## Find:

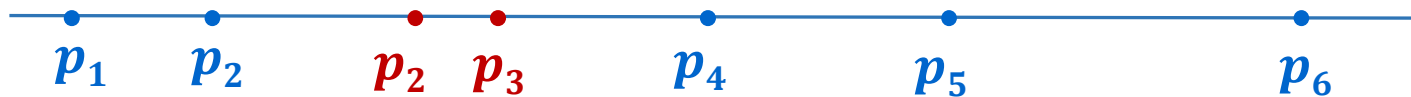
- A pair of points  $p_i, p_j$  s.t. the Euclidean distance  $d(p_i, p_j)$  is minimized

What is the first algorithm you can think of?

- Try all  $\Theta(n^2)$  possible pairs

Can we do better if dimension  $d = 1$  ?

# Closest Pair in 1 Dimension



Algorithm:

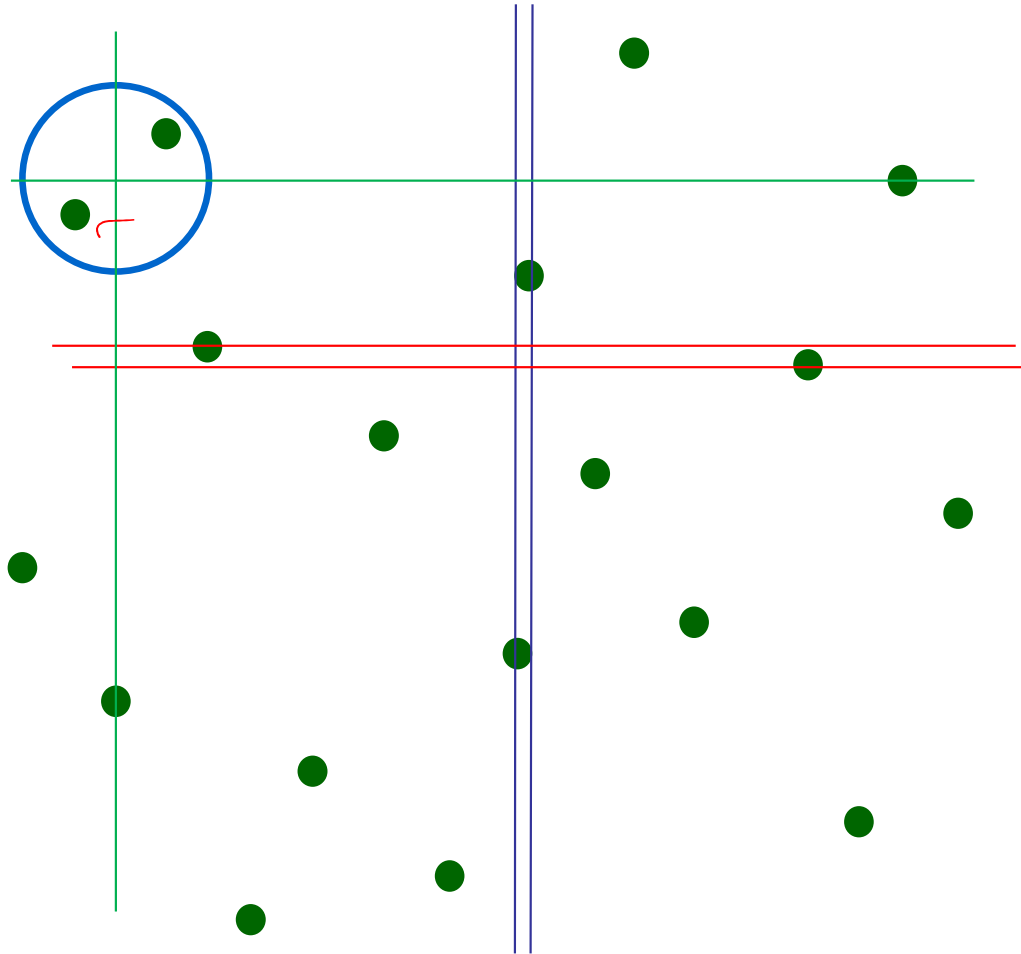
- Sort points so  $p_1 \leq p_2 \leq \dots \leq p_n$
- Find closest adjacent pair  $p_i, p_{i+1}$ .

Running time:  $O(n \log n)$

What about  $d = 2$  ?



# Closest Pair in 2 Dimensions



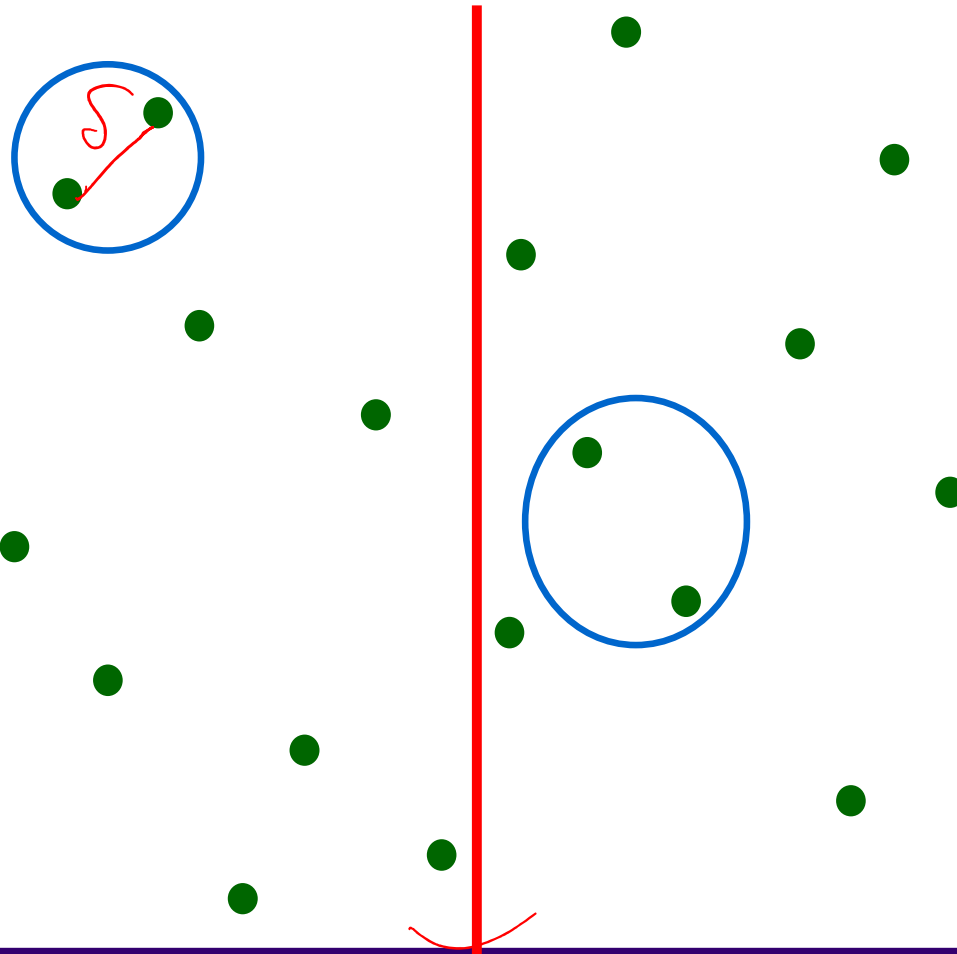
Sorting on 1<sup>st</sup> coordinate  
doesn't work

No single direction to sort  
points to guarantee success!

Let's try divide & conquer...

How might we divide the points so  
that each subpart is a constant  
factor smaller?

# Closest Pair in 2 Dimensions: Divide and Conquer



How might we divide the points so that each subpart is a constant factor smaller?

Split using **median  $x$ -coordinate!**

- each subpart has size  $n/2$ .

Conquer:

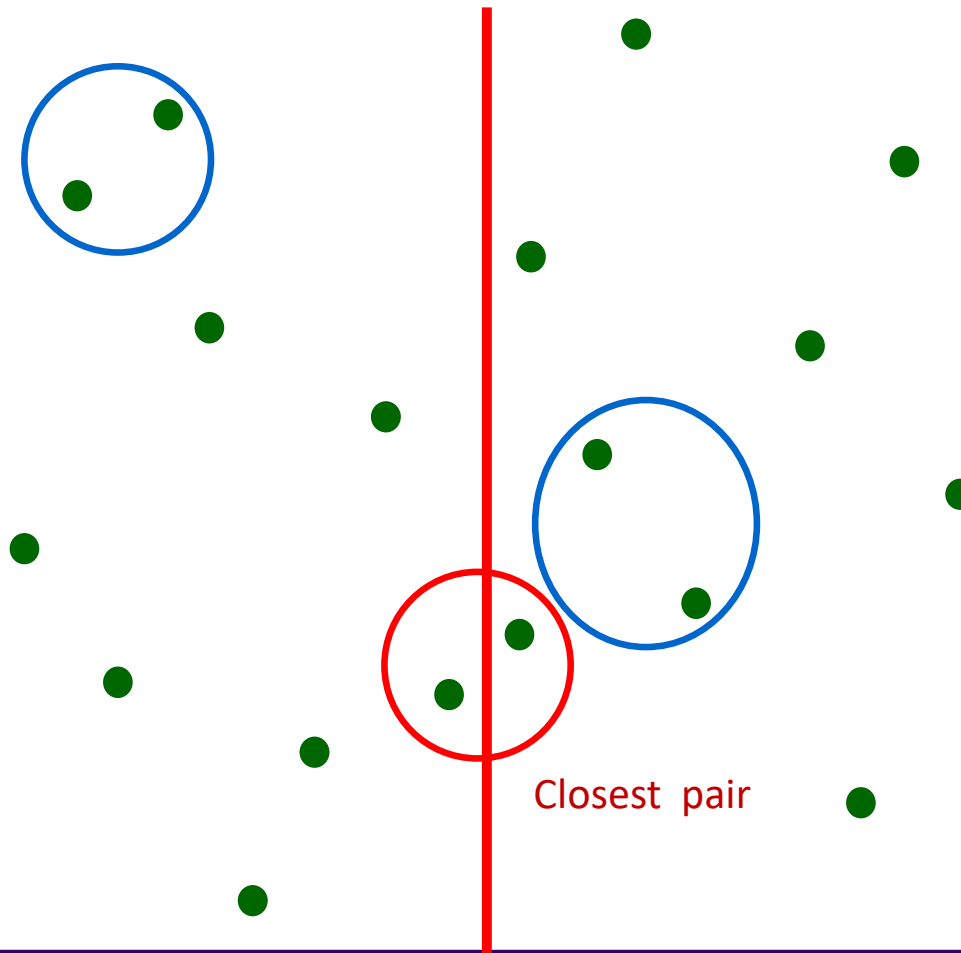
- Solve both size  $n/2$  subproblems recursively

Recombine to get overall answer?

Take the closer of the two answers?

- works here but....

# Closest Pair in 2 Dimensions: Divide and Conquer



How might we divide the points so that each subpart is a constant factor smaller?

Split using **median  $x$** -coordinate!

- each subpart has size  $n/2$ .

Conquer:

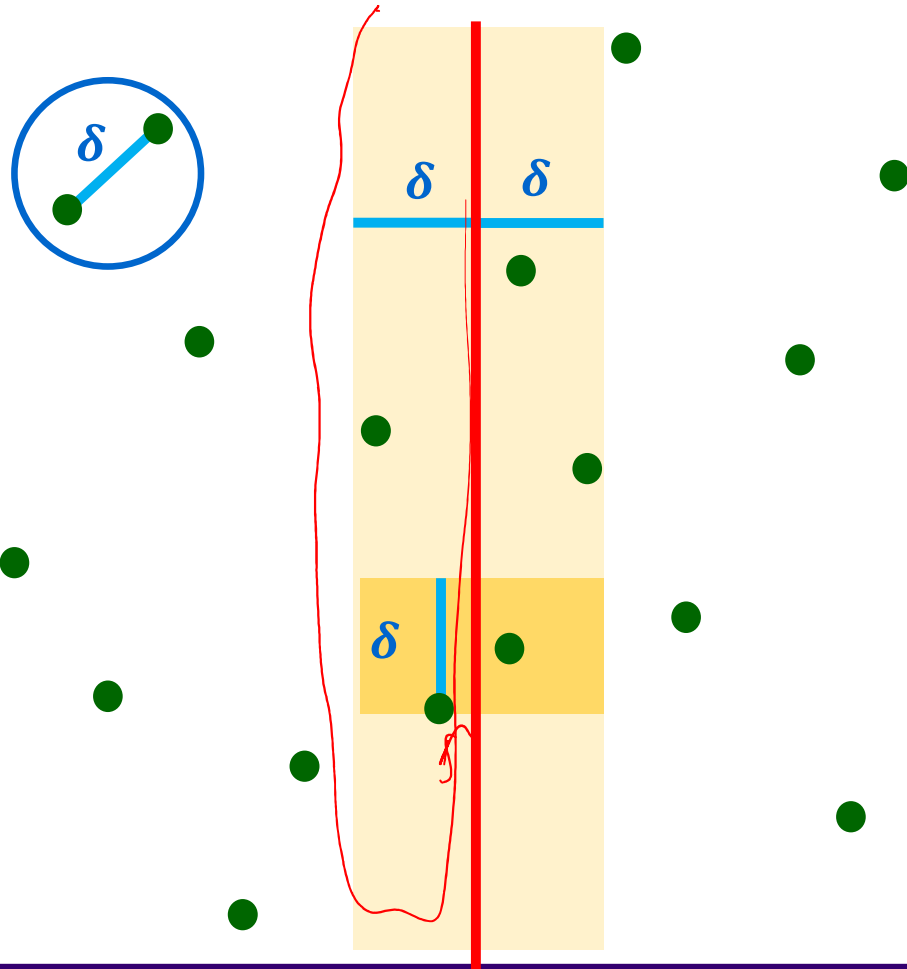
- Solve both size  $n/2$  subproblems recursively

Recombine to get overall answer?

Take the closer of the two answers?

- ...but not always!

# Closest Pair in 2 Dimensions: Divide and Conquer



Need to worry about pairs across the split!

New idea to handle them

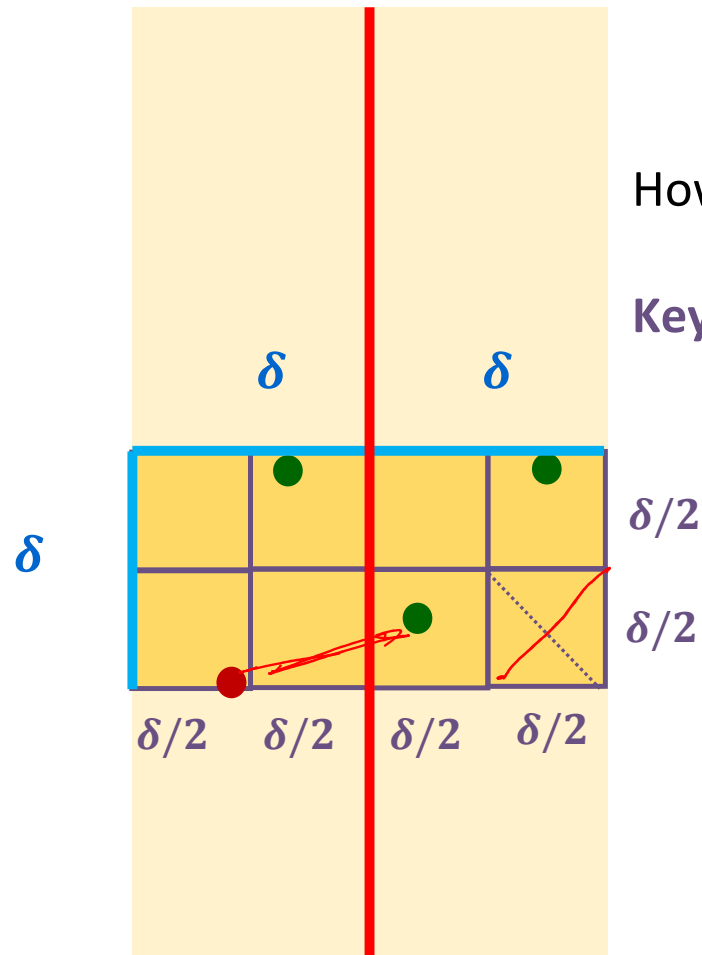
- Let  $\delta$  be the distance of the closest pair in the 2 subparts
- This pair is a candidate
- Only need to check width  $\delta$  band either side of the median

Within that band ...

- only need to compare each point with the other points in the rectangle of height  $\delta$  above it.

How many points can that be?

# Closest Pair in 2 Dimensions: Divide and Conquer



How many points can there be in that  $\delta$  by  $2\delta$  rectangle?

**Key idea:** We know that no pair on either side is closer than  $\delta$  apart so there can't be too many!

- Each of the 8 squares of side  $\delta/2$  can contain at most 1 point!
  - Because diagonal has length  $< \delta$
- So... only need to compare each point with the next 7 points above it to guarantee you'll find a partner closer than  $\delta$  in the rectangle if there is one!

# Closest Pair in 2 Dimensions: Divide and Conquer

## Fleshing out the algorithm:

### Divide:

- At top level we need median  $x$  coordinate to split points  $O(n \log n)$  total
- At next level down we'll need median  $x$  coordinate for each side over all calls
- Might as well sort all points by  $x$  coordinate up front to get all medians at once!

Conquer: Solve the two sub-problems to get two candidate pairs  $2 T(n/2)$

### Recombine:

- Choose closer candidate pair and let its distance be  $\delta$   $O(1)$
- Select  $B$  = all points in band with  $x$  coordinates within  $\delta$  of median  $O(n)$  ✓
- Sort  $B$  by  $y$  coordinate May involve repeated work for different calls  $O(n \log n)$  ←
- Compare each point in  $B$  with next 7 points and update if closer pair found.  $O(n)$  ✓

# Closest Pair in 2 Dimensions: Divide and Conquer

## Fleshing out the algorithm: A better version:

Preprocess: Compute sorted list  $X$  of points by  $x$  coordinate

- Subparts will be defined by two indices into this list

Compute sorted list  $Y$  of points by  $y$  coordinate

$O(n \log n)$

$O(n \log n)$

Divide: Use median in  $X$  to get  $X_L$  and  $X_R$  and filter points of  $Y$  to produce sorted sublists  $Y_L$  and  $Y_R$

$O(n)$

Conquer: Solve the two sub-problems to get two candidate pairs

$2T(n/2)$

Recombine:

- Choose closer candidate pair and let its distance be  $\delta$   $O(1)$
- Filter  $Y$  to get  $B$  = points in band w/  $x$  coordinates within  $\delta$  of median  $O(n)$
- Compare each point in  $B$  with next 7 points and update if closer pair found.  $O(n)$

# Closest Pair in 2 Dimensions: Divide and Conquer

Total runtime = Preprocessing time + Divide and Conquer time

Let  $T(n)$  be Divide and Conquer time:

Recurrence:

- $T(n) \leq 2T(n/2) + O(n)$  for  $n \geq 3$
- $T(2) = 1$

Solution:  $T(n)$  is  $O(n \log n)$ .

With preprocessing, total runtime is  $O(n \log n)$ .



# Sometimes two sub-problems aren't enough

## More general divide and conquer

$$T(n) = aT(n/b) + \underline{cn^k}$$

- You've broken the problem into  $a$  different sub-problems
- Each has size at most  $n/b$
- The cost of break-up and recombining sub-problem solutions is  $O(n^k)$ 
  - “cost at the top level”

## Recurrence

- $T(n) = a \cdot T(n/b) + O(n^k)$  for  $n \geq b$
- $T$  is constant for inputs  $< b$ .
  - For solutions correct up to constant factors no need for exact base case

# Solving Divide and Conquer Recurrence

Master Theorem: Suppose that  $T(n) = a \cdot T(n/b) + O(n^k)$  for  $n > b$ .

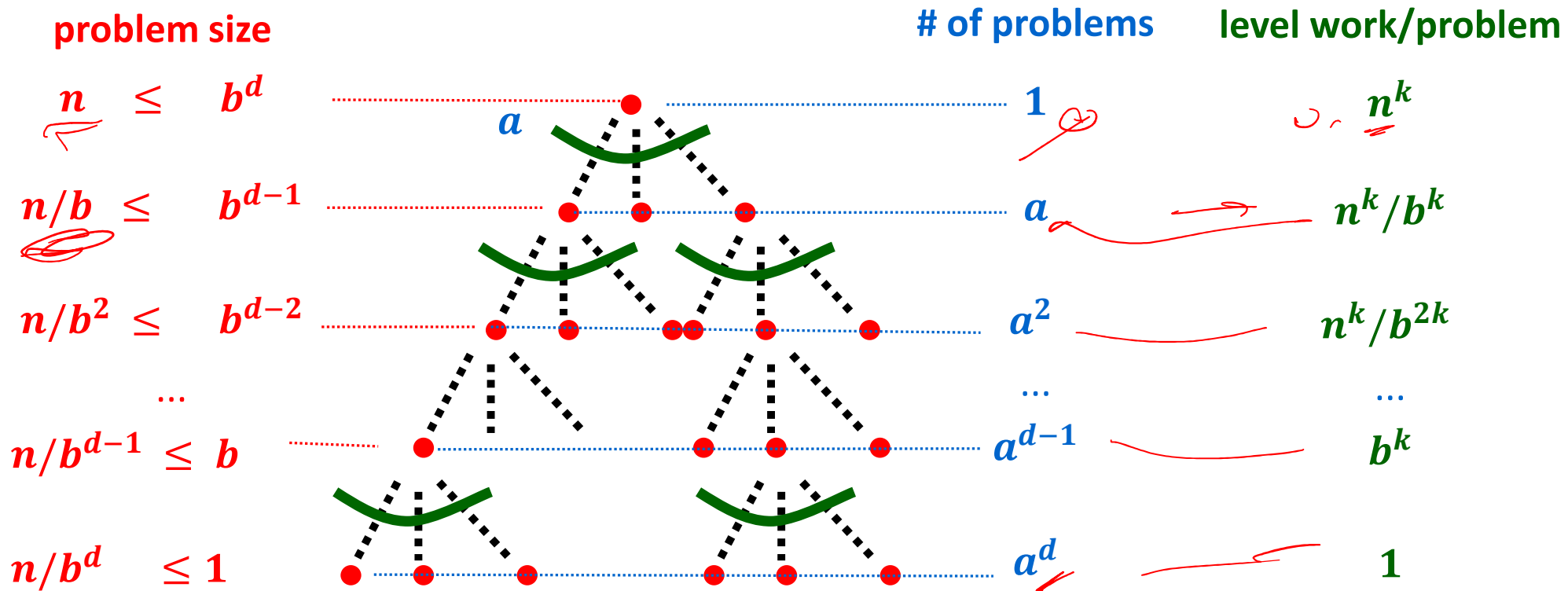
- If  $a < b^k$  then  $T(n)$  is  $O(n^k)$ 
  - Cost is dominated by work at top level of recursion
- • If  $a = b^k$  then  $T(n)$  is  $O(n^k \log n)$  ←
  - Total cost is the same for all  $\log_b n$  levels of recursion
- If  $a > b^k$  then  $T(n)$  is  $O(n^{\log_b a})$  ←
  - Note that  $\log_b a > k$  in this case
  - Cost is dominated by total work at lowest level of recursion

Binary search:  $a = 1, b = 2, k = 0$  so  $a = b^k$ : Solution:  $O(n^0 \log n) = O(\log n)$

Mergesort:  $a = 2, b = 2, k = 1$  so  $a = b^k$ : Solution:  $O(n^1 \log n) = O(n \log n)$

# Proving Master Theorem for $T(n) = a \cdot T(n/b) + c \cdot n^k$

Write  $d = \lceil \log_b n \rceil$  so  $n \leq b^d$



# Proving Master Theorem for $T(n) = a \cdot T(n/b) + c \cdot n^k$

Write  $d = \lceil \log_b n \rceil$  so  $n \leq b^d$

# of problems	level work/problem	total work/level
1	$n^k$	$a/b^k \cdot n^k$
$a$	$n^k/b^k$	$(a/b^k) \cdot n^k$
$a^2$	$n^k/b^{2k}$	$(a/b^k)^2 \cdot n^k$
...	...	...
$a^{d-1}$	$b^k$	...
$a^d$	1	$a^{\log_b n}$

**total work**

If  $a < b^k$  sum of geometric series with biggest term  $O(n^k)$

If  $a = b^k$  sum of  $O(\log n)$  terms each  $O(n^k)$

If  $a > b^k$  sum of geometric series with biggest term  $O(a^{\log_b n})$

**Claim:**  $a^{\log_b n} = n^{\log_b a}$

**Proof:** Take  $\log_b$  of both sides