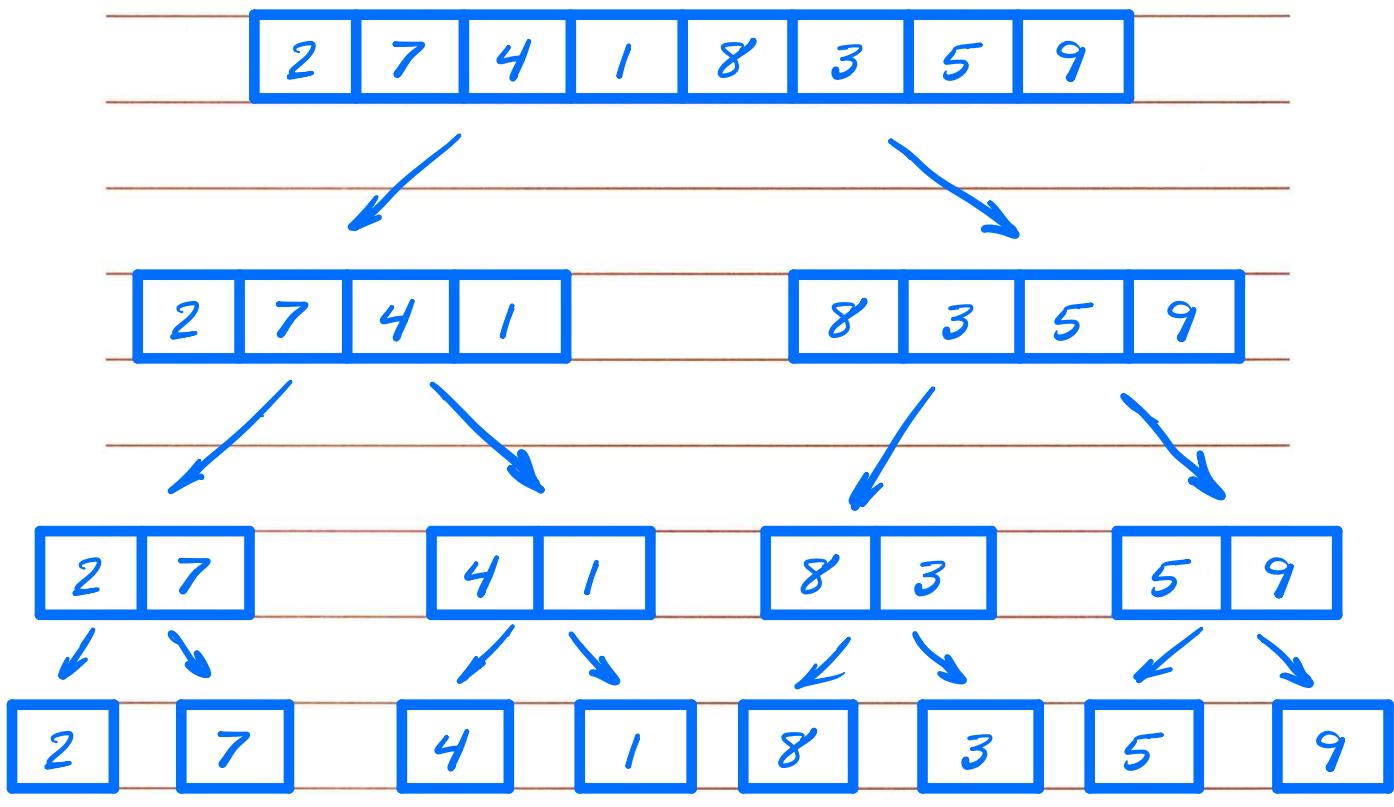


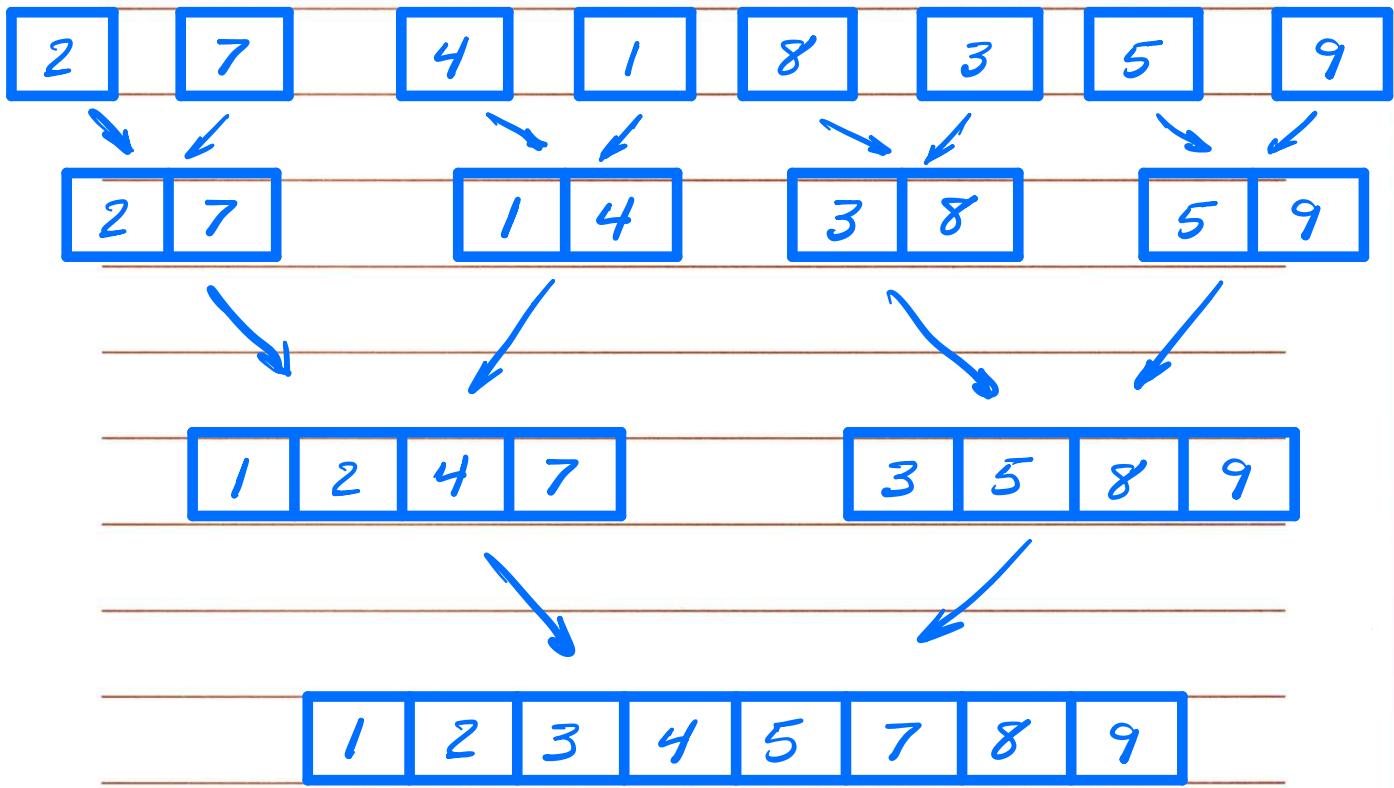
Divide & Conquer

1. Divide problem into n subproblems.

2. Conquer: i.e., solve the subproblems recursively, or if trivial solve the problem itself.

3. Combine the solutions to the subproblems.





MergeSort (A, p, r)

if p < r then

$$q = \lfloor (p+r)/2 \rfloor$$

MergeSort (A, p, q)

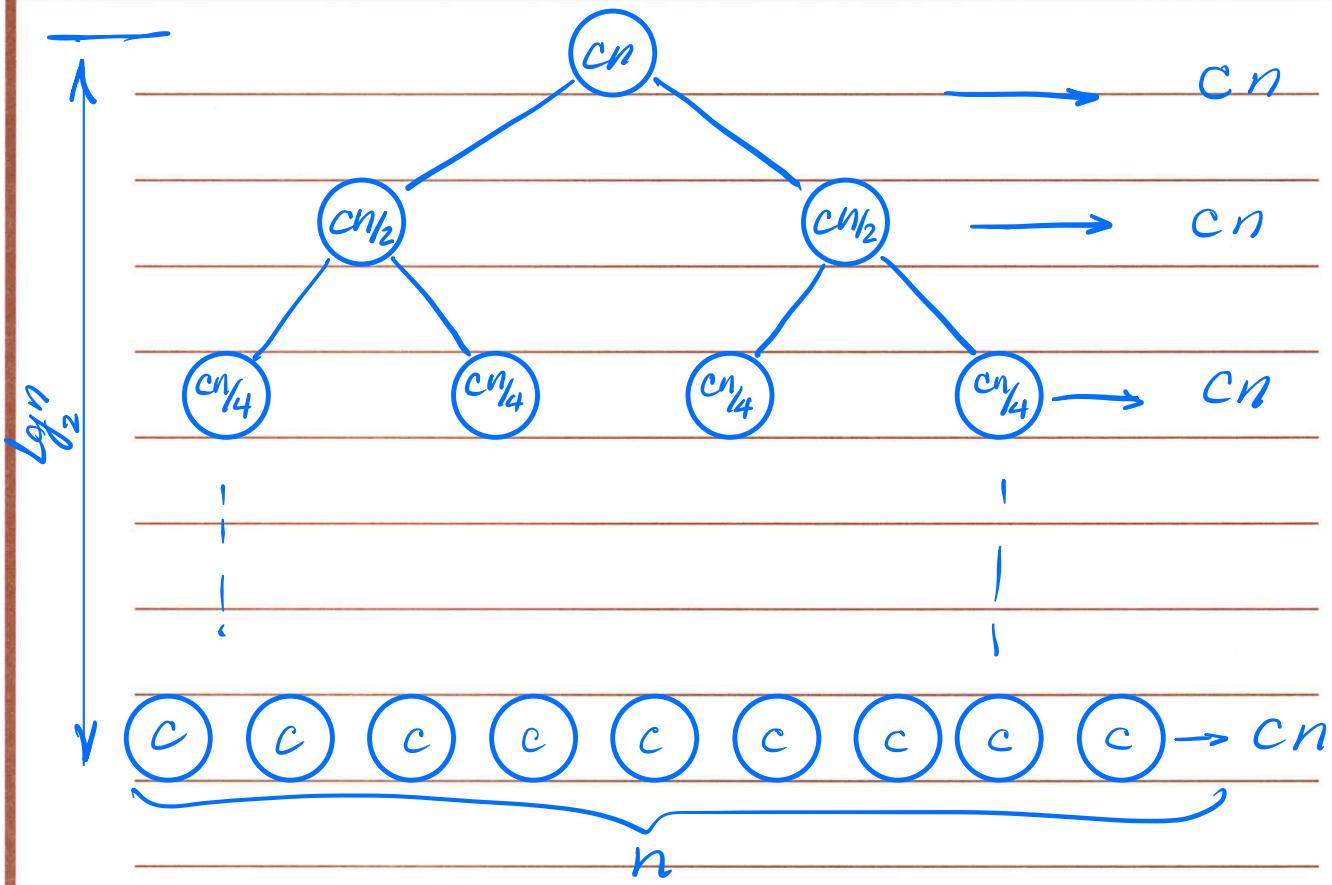
MergeSort (A, q+1, r)

Merge (A, p, q, r)

endif

Runtime Analysis: (Method 1)

- Draw the recursion tree
- Determine the cost of operations at each node of the tree.
- Determine the total cost of the algorithm by adding up the costs of all the nodes of the tree.



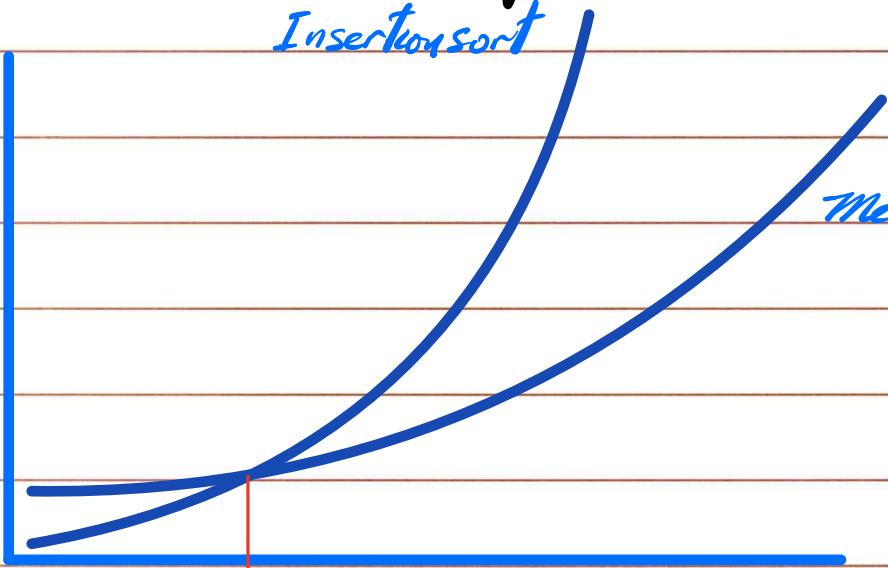
- Each level of the recursion tree takes c time
- There are $\log n$ levels in the tree.

$$\text{Total cost} = \sum_{i=1}^{\lg n} cn = cn \lg n = \Theta(n \lg n)$$

Merge sort - Insertion sort hybrid

Insertionsort

Mergesort



Proof of Correctness

Let $P(n)$ be the proposition that merge sort correctly sorts any array of length n .

Base Case:

Consider an array of 1 element. This array is already sorted.

∴ Base case proven

Inductive Step:

As inductive hypothesis we assume that $P(k)$ is true for all $1 \leq k < n$. We now need to show that merge sort also works correctly on any array of size n .

Suppose we call mergesort on an array of size n . It will then recursively call mergesort on two arrays of size $n/2$. By I.H. these calls will correctly sort these arrays.

Once the two halves are correctly sorted, we will use the merge function that correctly* combines the two sorted subarrays.

Therefore after the completion of the merge operation, we will have a correct sorting of the original array of size n .

*Note: In this case the POC for the combine step is trivial. If the combine step is non-trivial, it has to be proven separately.

This concludes the inductive step.
Having proven the base case and inductive step we have proven that $P(n)$ is true for all $n \geq 1$

Routine Analysis: (Method 2)

Divide - Takes $O(1)$

Conquer - If the original problem takes $T(n)$ time, the two subproblems take
 $2 \cdot T(n/2)$

Combine - Takes $O(n)$ on array of size n

Recurrence Eq. for merge sort:

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

Conquer Combine Divide

Our recurrence equation for a D&C solution will (usually) look like:

$$T(n) = \begin{cases} O(1) & \text{if } n \leq c \\ a \cdot T\left(\frac{n}{b}\right) + D(n) + C(n) & \text{otherwise} \end{cases}$$

Annotations:

- # of subproblems: points to the term a in $a \cdot T\left(\frac{n}{b}\right)$
- size of each subproblem: points to the term $\frac{n}{b}$ in $T\left(\frac{n}{b}\right)$
- time to divide: points to the term $D(n)$
- time to combine: points to the term $C(n)$

Master Method

It is a cookbook method for solving recurrences of the form

$$T(n) = a T\left(\frac{n}{b}\right) + f(n)$$

- where $a \geq 1, b \geq 1$ are constants
- $f(n)$ is an asymptotically positive function

Master Theorem

Given the above definitions of the recurrence relation, $T(n)$ can be bounded asymptotically as follows:

1- If $f(n) = O(n^{b - \epsilon})$ for some $\epsilon > 0$,

$$\text{then } T(n) = \Theta(n^b)$$

2- If $f(n) = \Theta(n^b)$ then

$$T(n) = \Theta(n^b \lg n)$$

3- If $f(n) = \Omega(n^{b + \epsilon})$ for some $\epsilon > 0$,

and if $a f(n/b) \leq c f(n)$ for some constant $c < 1$ and all

sufficiently large n , then

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

Ex.

$$\left\{ \begin{array}{l} n^{\frac{\log a}{b}} = n \\ f(n) = n \lg n \end{array} \right.$$

Does this fall into case #3?

No! $f(n)$ does grow faster than $n^{\frac{\log a}{b}}$,
but only logarithmically.

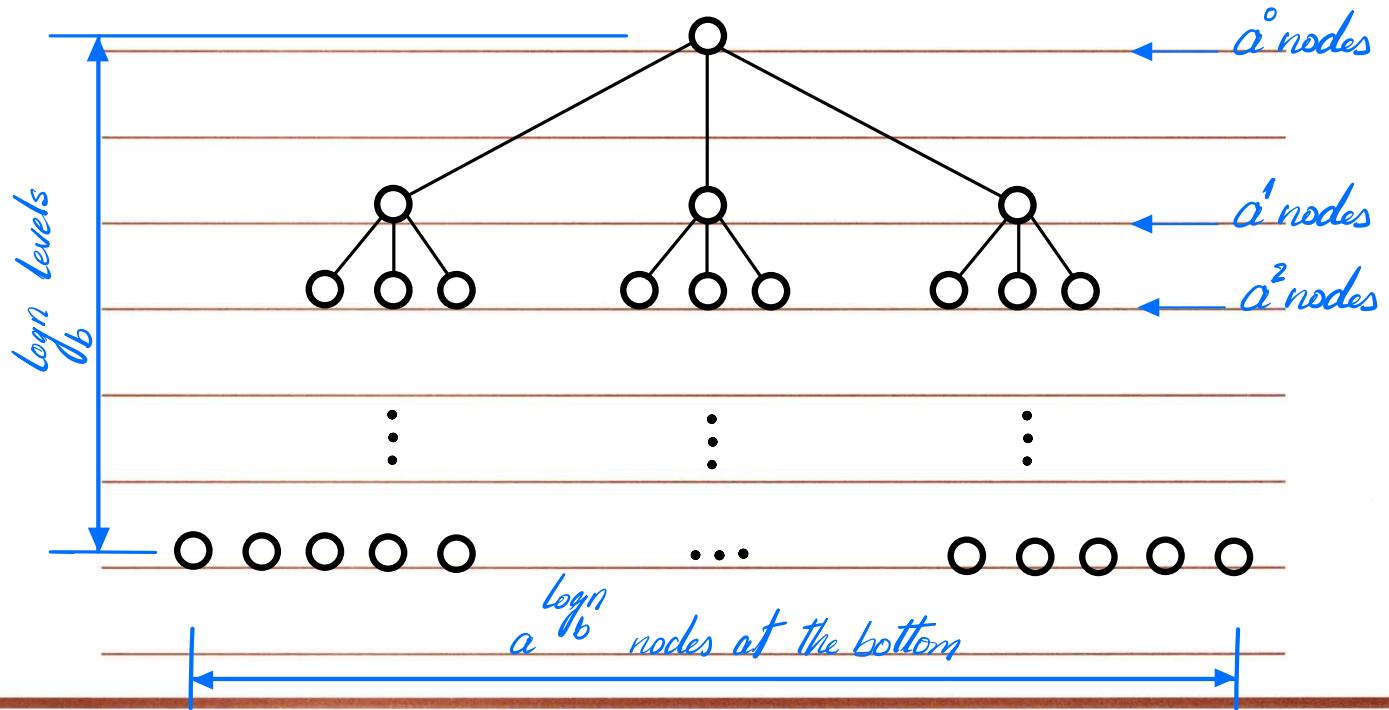
Generalizing case #2 of the Master Method

2- If $f(n) = \Theta(n^{\frac{\log a}{b}} \lg^k n)$ where $k \geq 0$,

then $T(n) = \Theta(n^{\frac{\log a}{b}} \lg^{k+1} n)$

Intuition Behind the Master Method

Case #1



$$a^{\log_b n} = n^{\log_b a}$$

total no. of subproblems =

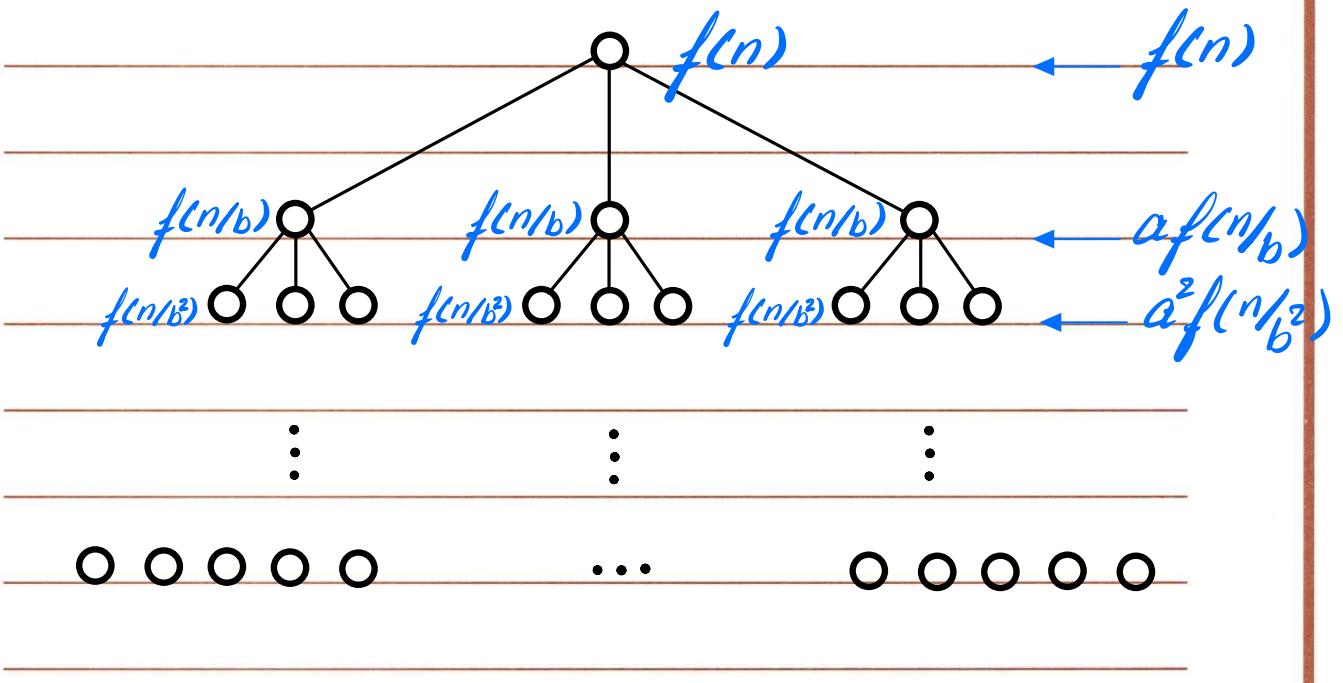
$$n^{\log_b a} + \frac{1}{a} n^{\log_b a} + \frac{1}{a^2} n^{\log_b a} + \dots$$

$$= \frac{a}{(a-1)} n^{\log_b a} = \Theta(n^{\log_b a})$$

In case #1, the no. of subproblems dominates the complexity. Here, the tree is bottom heavy.

Intuition Behind the Master Method

Case #3



Regularity condition:

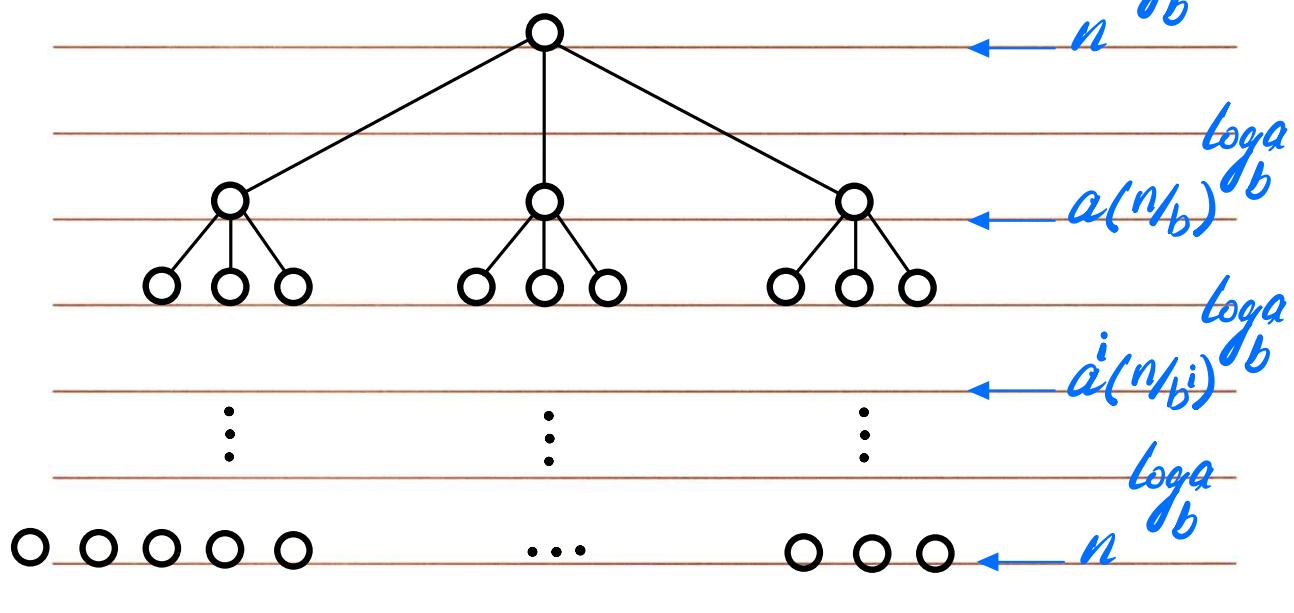
if $a f(n/b) \leq c f(n)$ for some $c < 1$,
then the cost of the combine and divide
steps at each level must be a fraction of
that in the previous level.

$$\text{So, } f(n) + C f(n) + C^2 f(n) + \dots = \\ \frac{f(n)}{1/(1-C)} = \Theta(f(n))$$

In case #3, the cost of combine and divide steps
dominate the complexity. Here, the tree is top heavy.

Intuition Behind the Master Method

Case #2 $f(n) = \Theta(n^{\log_b a})$



$$\text{At level } i, a^i (n/b)^{\log_b a} = n^{\log_b a} (a^i \cdot b^{-i \log_b a})$$

$$= n^{\log_b a} (a^i (b^{\log_b a})^{-i})$$

$$= n^{\log_b a} (a^i \cdot a^{-i}) = n^{\log_b a}$$

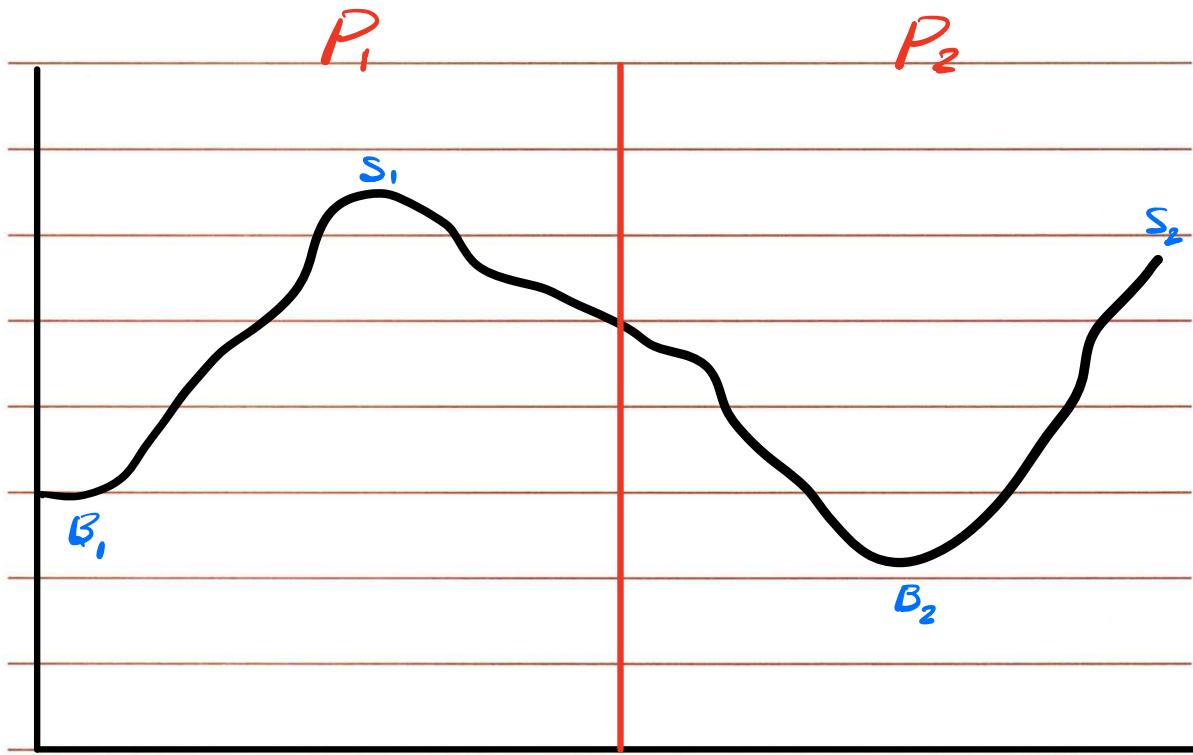
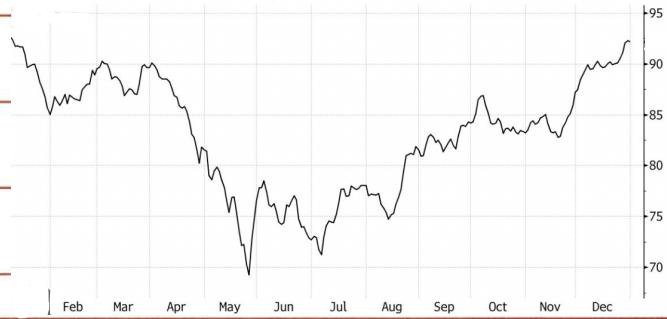
Since we have $\lg n$ levels, $T(n) = \Theta(n^{\log_b a} \lg n)$

In case #2, the cost of operations at each level is the same. The tree levels are equally weighted

Stock Market Problem

Given the price chart for a stock over a period of 1 day, which buy and sell days give us the maximum profit? (must buy before selling)

Brute force approach takes $\Theta(n^2)$ time.



Solve P_1 & P_2 recursively to find B_1, S_1, B_2 and S_2

Merge Step : Three cases to consider

Case 1 - Buy and Sell in P_1

$$B = B_1$$

$$S = S_1$$

Case 2 - Buy and Sell in P_2

$$B = B_2$$

$$S = S_2$$

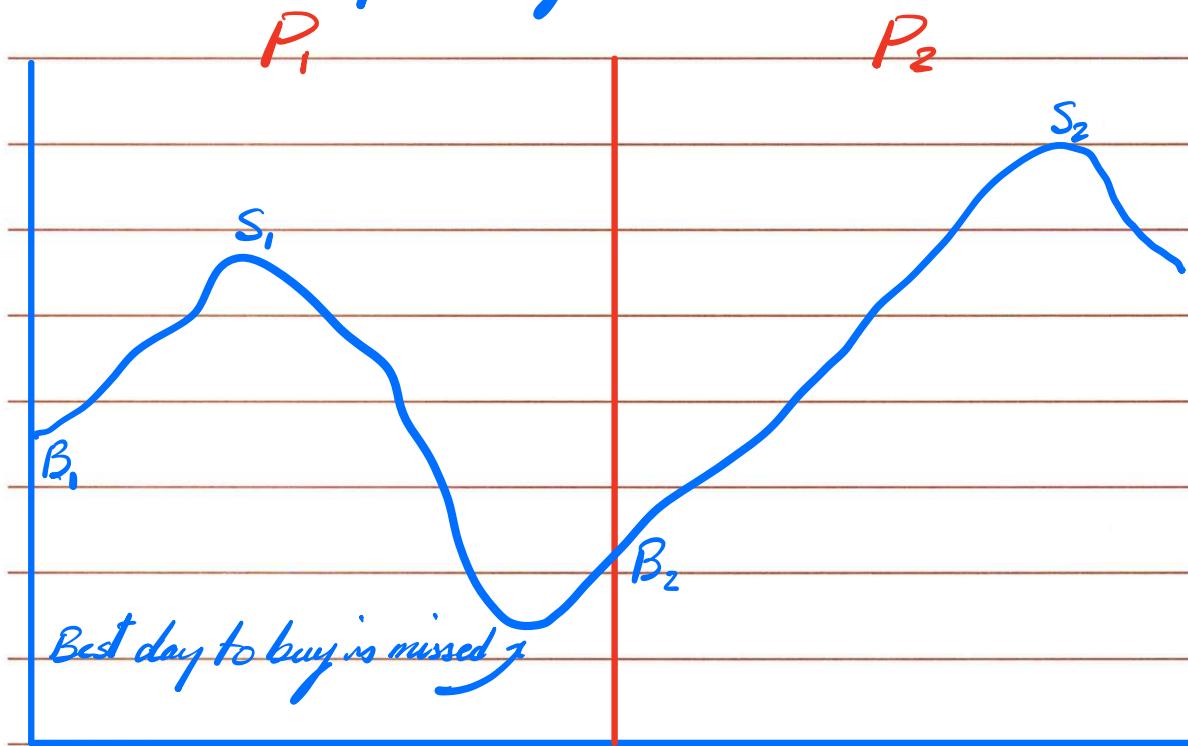
Case 3 - Buy in P_1 and Sell in P_2

$$B = B_1$$

$$S = S_2$$

This may not
always work!

Consider the following case



Corrected Merge Step

$$M = \min(M_1, M_2)$$

$$X = \max(X_1, X_2)$$

Case 1 - Buy and Sell in P_1

$$B = B_1, S = S_1$$

Case 2 - Buy and Sell in P_2

$$B = B_2, S = S_2$$

Case 3 - Buy in P_1 and Sell in P_2

$$B = M_1, S = X_2$$

Complexity analysis - Using The Master Theorem

$$a=2, b=2, \text{ and}$$

The cost of the divide and combine steps = $\Theta(1)$

$$n^{\frac{\log a}{b}} = n^{\frac{\log 2}{2}} = n^1$$

$$f(n) = \Theta(1)$$

\Rightarrow Case 1:

$$T(n) = \Theta(n)$$

Dense Matrix Multiplication

Assuming square dense matrices.

$$n \left\{ \underbrace{\begin{bmatrix} A \end{bmatrix}}_n \underbrace{\begin{bmatrix} B \end{bmatrix}}_n = \underbrace{\begin{bmatrix} C \end{bmatrix}}_n \right\}_n$$

Brute force approach leads to $T(n) = \Theta(n^3)$

Try divide & conquer

$$\begin{bmatrix} A_{11} & A_{12} \\ \hline A_{21} & A_{22} \end{bmatrix} \quad \begin{bmatrix} B_{11} & B_{12} \\ \hline B_{21} & B_{22} \end{bmatrix} \quad \begin{bmatrix} C_{11} & C_{12} \\ \hline C_{21} & C_{22} \end{bmatrix}$$

$$C_{11} = A_{11} \cdot B_{11} + A_{12} \cdot B_{21}$$

$$C_{12} = A_{11} \cdot B_{12} + A_{12} \cdot B_{22}$$

$$C_{21} = A_{21} \cdot B_{11} + A_{22} \cdot B_{21}$$

$$C_{22} = A_{21} \cdot B_{12} + A_{22} \cdot B_{22}$$

Complexity analysis - Using The Master Theorem

$a=8, b=2$, and

matrix additions

The cost of the $\underbrace{\text{divide}}_{\Theta(1)}$ and $\underbrace{\text{combine}}_{\Theta(n^2)}$ steps = $\Theta(n^2)$

$$n^{\log_b a} = n^{\log_2 8} = n^3$$

$$f(n) = \Theta(n^2)$$

\Rightarrow Case 1:

$$T(n) = \Theta(n^3)$$

Compute ? $n/2 \times n/2$ intermediate matrices
(Strassen's Algorithm)

$$P = (A_{11} + A_{22})(B_{11} + B_{22})$$

$$Q = (A_{21} + A_{22})B_{11}$$

$$R = A_{11}(B_{12} - B_{22})$$

$$S = A_{22}(B_{21} - B_{11})$$

$$T = (A_{11} + A_{12})B_{22}$$

$$U = (A_{21} - A_{11})(B_{11} + B_{12})$$

$$V = (A_{12} - A_{22})(B_{21} + B_{22})$$

$$C_{11} = P + S - T + V$$

$$C_{12} = R + T$$

$$C_{21} = Q + S$$

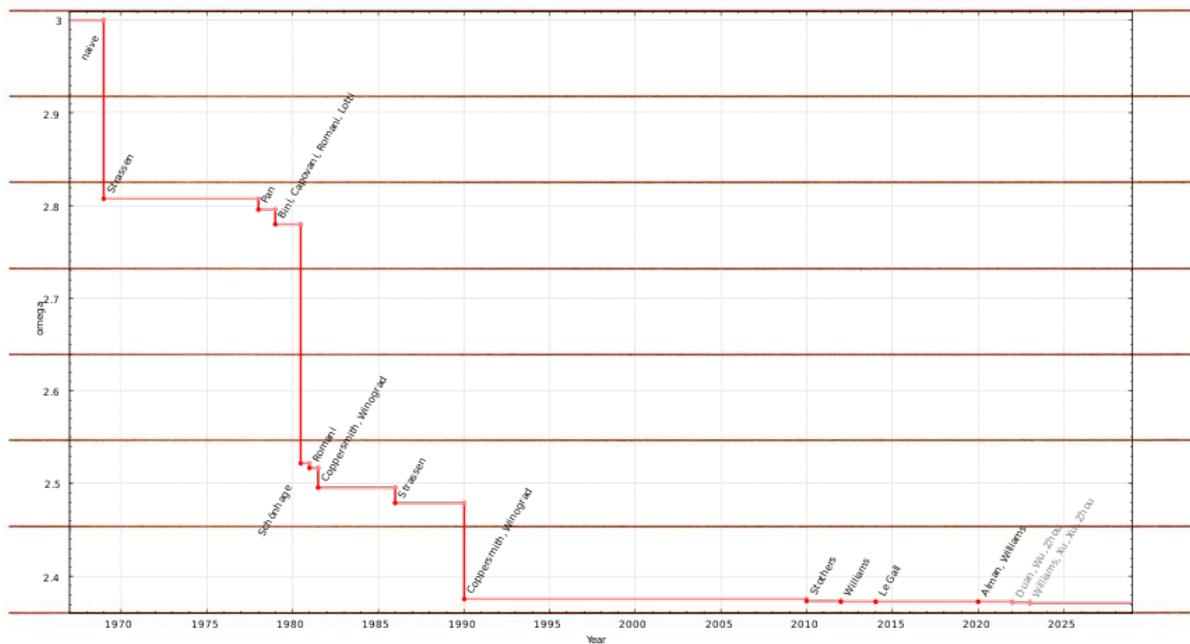
$$C_{22} = P + R - Q + U$$

Complexity Analysis

$$a=7, b=2, f(n)=\Theta(n^2)$$

$$n^{\frac{\log a}{b}} = n^{\frac{\log 7}{2}} \approx n^{2.81}$$

$$\Rightarrow \text{Case } \#1: T(n) = \Theta(n^{2.81})$$

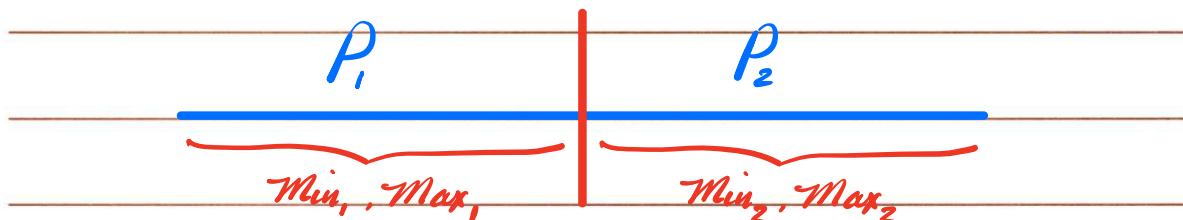


From Wikipedia

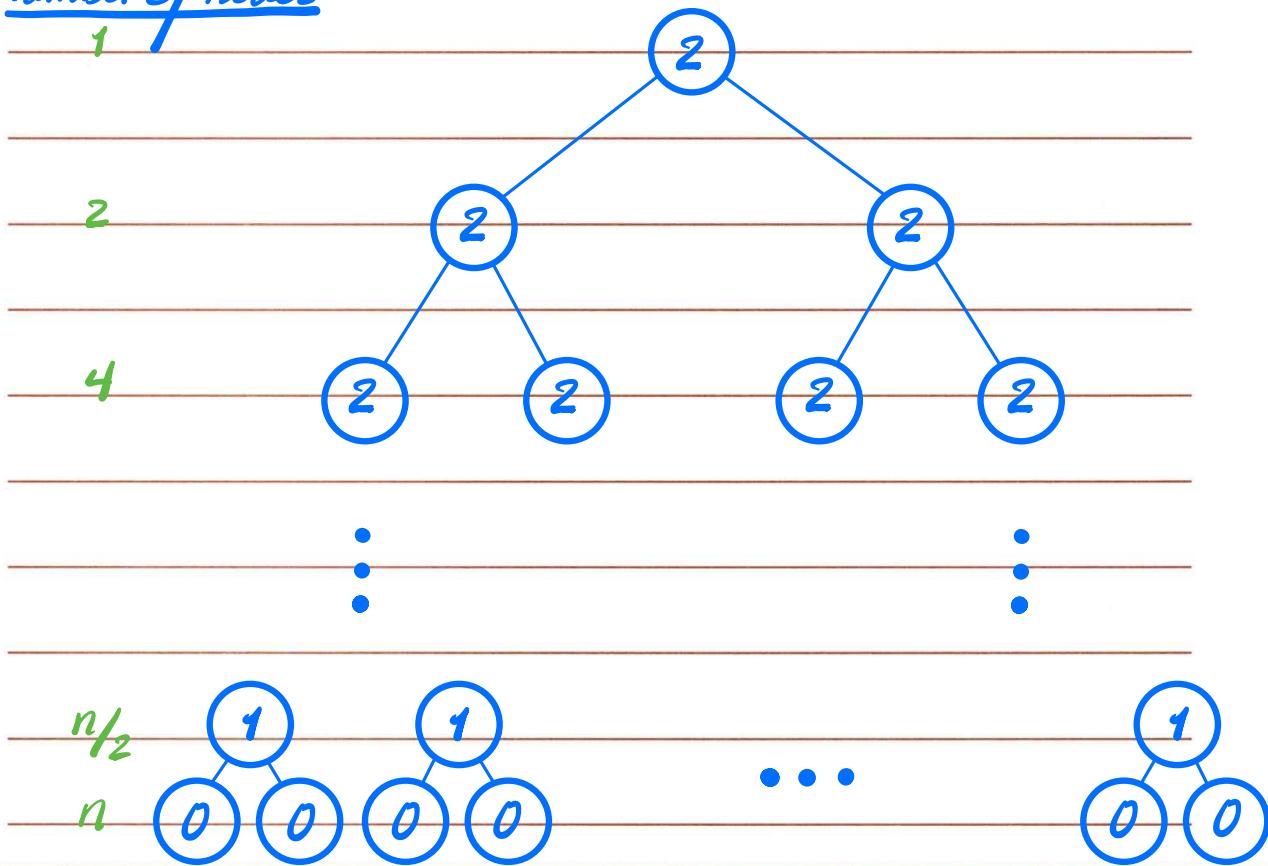
Finding the minimum and the maximum in an unsorted array.

Number of comparisons required:

- Brute force $(n-1) + (n-1) = 2n - 2 = \Theta(n)$
- Divide & Conquer ?



Number of nodes



Total no. of nodes above the bottom level = $n-1$

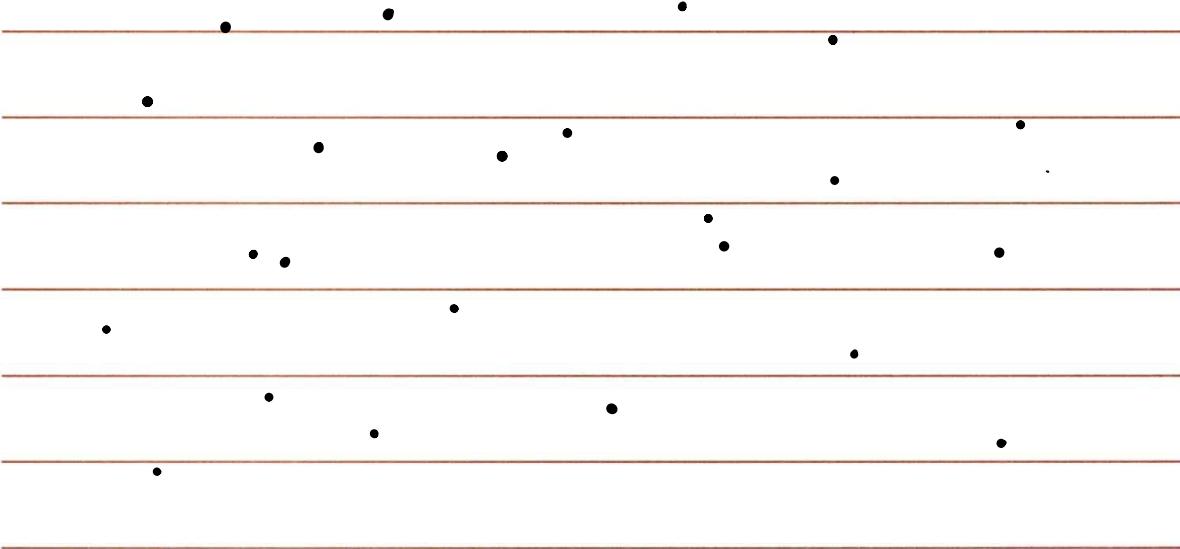
total no. of comparisons required =

$$(n-1) \times 2 - n/2$$

$$= 3n/2 - 2$$

Problem Statement

Find the closest pair of points on a Euclidean plane



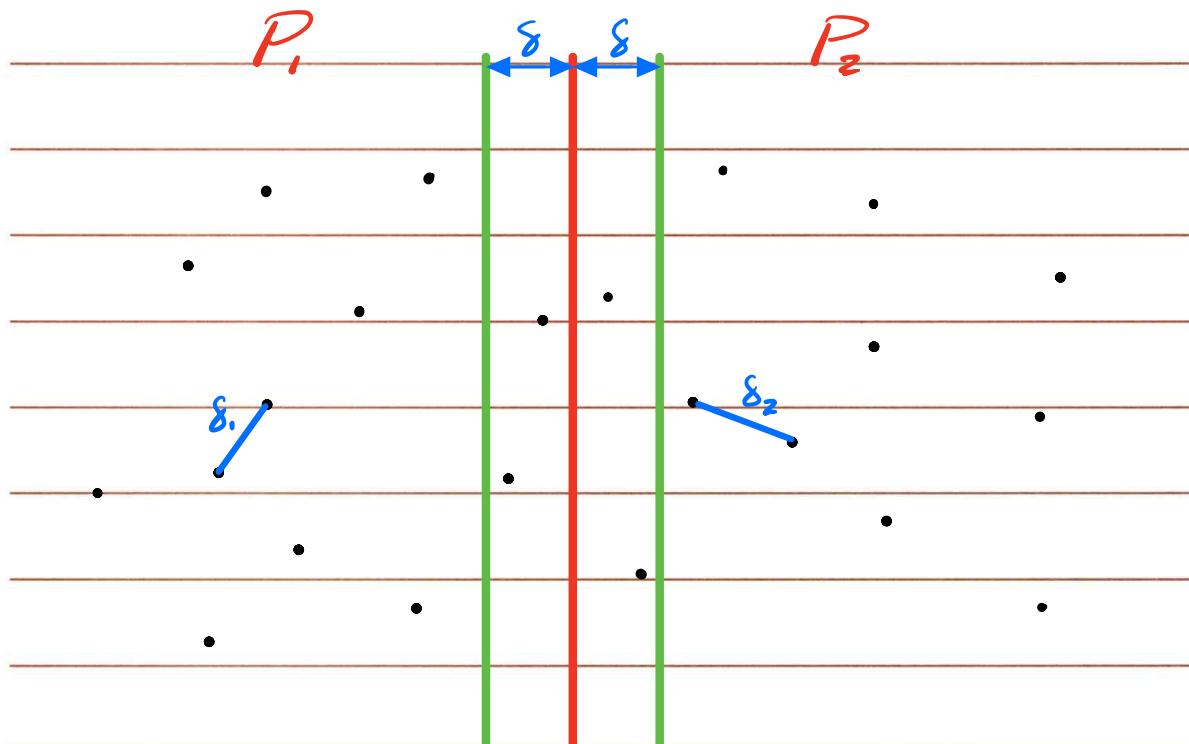
Number of distance calculations and comparisons required:

- Brute force

$$(n-1) + (n-2) + \dots + 1 = \Theta(n^2)$$

- Divide & Conquer

?



$$\delta = \min(\delta_1, \delta_2)$$

Merge Step : Three cases to consider

Case 1 - Both points are in P_1

(found recursively)

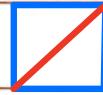
Case 2 - Both points are in P_2

(found recursively)

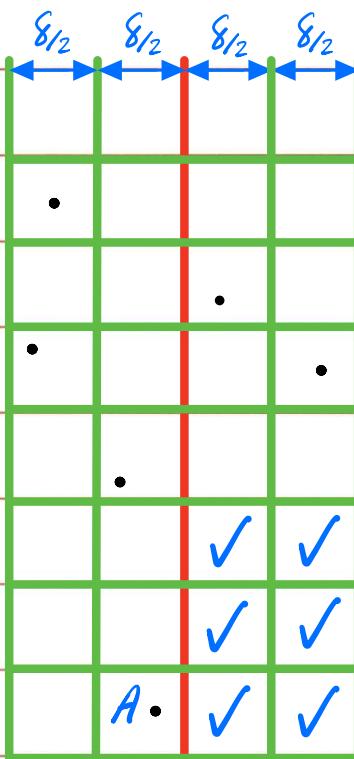
Case 3 - One point is in P_1 and the other in P_2

How do we find these two pts.?

Consider a $8_{1/2} \times 8_{1/2}$ square



Farthest two points within this square
are $\frac{8\sqrt{2}}{2}$ away from each other. ($\frac{8\sqrt{2}}{2} < 8$)



✓ candidate cells
to look for a pt.
that is closer than
8 to A.

Implementations

closest-pair (P)

$\Theta(n \lg n)$ Construct P_x : List of pts. sorted by x -coord.

$\Theta(n \lg n)$ Construct P_y : List of pts. sorted by y -coord.

$(p_0, p_1) = \text{closest-pair-Rec}(P_x, P_y)$

closest-pair-Rec(P_x, P_y)

if $|P| \leq 3$ then

solve it directly

else

$\Theta(1)$ construct Q_x ... left half of P_x

$\Theta(n)$ construct Q_y ... list of points in Q_x sorted by y -coord.

$\Theta(1)$ construct R_x ... right half of P_x

$\Theta(n)$ construct R_y ... list of points in R_x sorted by y -coord.

$$(q_0, q_1) = \text{closest-pair-Rec}(Q_x, Q_y)$$

$$(r_0, r_1) = \text{closest-pair-Rec}(R_x, R_y)$$

$$\Theta(1) \quad \delta = \min(d(q_0, q_1), d(r_0, r_1))$$

$\Theta(n)$ {
 S = set of points in P within distance
 of δ from L (the center line)
 Construct S_y ... set of points in S sorted
 by y -coord.

$\Theta(n)$ {
 for each point $s \in S_y$, compute distance
 from s to each of the next 11 points in S_y
 Let (s, s') be the pair with min. distance
 if $d(s, s') < \delta$ then
 Return (s, s')
 elseif $d(q_0, q_1) < d(r_0, r_1)$ then
 Return (q_0, q_1)
 else
 Return (r_0, r_1)
 endif

Complexity analysis - Using The Master Theorem

$a=2$, $b=2$, and

The cost of the divide and combine steps = $\Theta(n)$

$$n^{\frac{\log a}{b}} = n^{\frac{\log 2}{2}} = n^1$$

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

→ Case 2:

$$T(n) = \Theta(n \lg n)$$

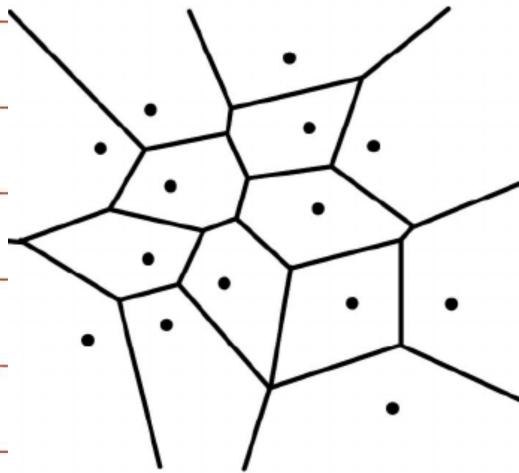
Cost of the preparation step (driver) : $\Theta(n \lg n)$

Cost of The divide & conquer solution: $\Theta(n \lg n)$

Overall worst-case time complexity: $\Theta(n \lg n)$

All Nearest Neighbors Problem

Given n points in the plane, find a nearest neighbor of each



Voronoi Diagrams

