# **CSE 421**

# Divide and Conquer: Finding Root Closest Pair of Points

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#### Finding the Closest Pair of Points

# A Divide and Conquer Alg

Divide: draw vertical line L with ≈ n/2 points on each side.
Conquer: find closest pair on each side, recursively.
Combine to find closest pair overall

**Return best solutions** 





# **Key Observation**

Suppose  $\delta$  is the minimum distance of all pairs in left/right of L.  $\delta = \min(12,21) = 12.$ 

Key Observation: suffices to consider points within  $\delta$  of line L.

Almost the one-D problem again: Sort points in 2δ-strip by their y coordinate.



### Almost 1D Problem

Partition each side of L into  $\frac{\delta}{2} \times \frac{\delta}{2}$  squares

Claim: No two points lie in the same  $\frac{\delta}{2} \times \frac{\delta}{2}$  box. Pf: Such points would be within

$$\sqrt{\left(\frac{\delta}{2}\right)^2 + \left(\frac{\delta}{2}\right)^2} = \delta \sqrt{\frac{1}{2}} \approx 0.7\delta < \delta$$

Let  $s_i$  have the i<sup>th</sup> smallest y-coordinate among points in the  $2\delta$ -width-strip.

Claim: If |i - j| > 11, then the distance between  $s_i$  and  $s_j$  is  $> \delta$ . Pf: only 11 boxes within  $\delta$  of  $y(s_i)$ .



## Recap: Finding Closest Pair



So, enough to check distance Distance of 30 to 19...41.

## Closest Pair (2Dim Algorithm)

```
Closest-Pair(p<sub>1</sub>, ..., p<sub>n</sub>) {
    if(n <= ??) return ??</pre>
```

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

 $\delta_1$  = Closest-Pair(left half)  $\delta_2$  = Closest-Pair(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.

return  $\delta$ .

}

# **Closest Pair Analysis I**

Let D(n) be the number of pairwise distance calculations in the Closest-Pair Algorithm when run on  $n \ge 1$  points

$$D(n) \leq \begin{cases} 1 & \text{if } n = 1\\ 2D\left(\frac{n}{2}\right) + 11n & \text{o.w.} \end{cases} \Rightarrow D(n) = \Theta(n\log n)$$

BUT, that's only the number of distance calculations What if we counted running time?

$$T(n) \leq \begin{cases} 1 & \text{if } n = 1\\ 2T\left(\frac{n}{2}\right) + O(n\log n) & \text{o.w.} \end{cases} \Rightarrow D(n) = \Theta(n\log^2 n)$$

## Can we do better? (Analysis II)

Yes!!

Don't sort by y-coordinates each time.

Sort by x at top level only.

This is enough to divide into two equal subproblems in O(n) Each recursive call returns  $\delta$  and list of all points sorted by y Sort points by y-coordinate by merging two pre-sorted lists.

$$T(n) \leq \begin{cases} 1 & \text{if } n = 1\\ 2T\left(\frac{n}{2}\right) + O(n) & \text{o.w.} \end{cases} \Rightarrow D(n) = \Theta(n \log n)$$

#### **Master Theorem**

Suppose  $T(n) = a T\left(\frac{n}{b}\right) + cn^k$  for all n > b. Then,

• If 
$$a > b^k$$
 then  $T(n) = \Theta(n^{\log_b a})$ 

• If 
$$a < b^k$$
 then  $T(n) = \Theta(n^k)$ 

• If 
$$a = b^k$$
 then  $T(n) = \Theta(n^k \log n)$ 

Works even if it is  $\left|\frac{n}{b}\right|$  instead of  $\frac{n}{b}$ . We also need  $a \ge 1, b > 1$ ,  $k \ge 0$  and T(n) = O(1) for  $n \le b$ .

#### **Master Theorem**

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$$a = b^k$$
 then  $T(n) = \Theta(n^k \log n)$ 

Example: For mergesort algorithm we have  $T(n) = 2T\left(\frac{n}{2}\right) + O(n).$ So,  $k = 1, a = b^k$  and  $T(n) = \Theta(n \log n)$ 

#### **Proving Master Theorem**



#### A Useful Identity

Theorem: 
$$1 + x + x^2 + \dots + x^d = \frac{x^{d+1} - 1}{x - 1}$$

Pf: Let  $S = 1 + x + x^2 + \dots + x^d$ 

Then,  $xS = x + x^2 + \dots + x^{d+1}$ 

So, 
$$xS - S = x^{d+1} - 1$$
  
i.e.,  $S(x - 1) = x^{d+1} - 1$   
Therefore,

$$S = \frac{x^{d+1} - 1}{x - 1}$$

Solve:  $T(n) = aT\left(\frac{n}{b}\right) + cn^k$ ,  $a > b^k$ 



Solve: 
$$T(n) = aT\left(\frac{n}{b}\right) + cn^k$$
,  $a = b^k$ 

$$T(n) = cn^k \sum_{i=0}^{\log_b n} \left(\frac{a}{b^k}\right)^i$$
$$= cn^k \log_b n$$

#### **Master Theorem**

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Works even if it is  $\left|\frac{n}{b}\right|$  instead of  $\frac{n}{b}$ . We also need  $a \ge 1, b > 1$ ,  $k \ge 0$  and T(n) = O(1) for  $n \le b$ .

# **Integer Multiplication**

# **Integer Arithmetic**



# How to use Divide and Conquer?

Suppose we want to multiply two 2-digit integers (32,45). We can do this by multiplying four 1-digit integers Then, use add/shift to obtain the result:

$$x = 10x_1 + x_0$$
  

$$y = 10y_1 + y_0$$
  

$$xy = (10x_1 + x_0)(10y_1 + y_0)$$
  

$$= 100 x_1y_1 + 10(x_1y_0 + x_0y_1) + x_0y_0$$

Same idea works when multiplying n-digit integers:

- Divide into 4 n/2-digit integers.
- Recursively multiply
- Then merge solutions



### A Divide and Conquer for Integer Mult

Let *x*, *y* be two n-bit integers Write  $x = 2^{n/2}x_1 + x_0$  and  $y = 2^{n/2}y_1 + y_0$ where  $x_0, x_1, y_0, y_1$  are all n/2-bit integers.

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

Therefore,

$$T(n) = 4T\left(\frac{n}{2}\right) + \Theta(n)$$

We only need 3 values  $x_1y_1, x_0y_0, x_1y_0 + x_0y_1$ Can we find all 3 by only 3 multiplication?

So,

 $T(n) = \Theta(n^2).$ 

### Key Trick: 4 multiplies at the price of 3

$$x = 2^{n/2} \cdot x_1 + x_0$$
  

$$y = 2^{n/2} \cdot y_1 + y_0$$
  

$$xy = (2^{n/2} \cdot x_1 + x_0)(2^{n/2} \cdot y_1 + y_0)$$
  

$$= 2^n \cdot x_1 y_1 + 2^{n/2} (x_1 y_0 + x_0 y_1) + x_0 y_0$$
  

$$\alpha = x_1 + x_0$$
  

$$\beta = y_1 + y_0$$
  

$$\alpha \beta = (x_1 + x_0)(y_1 + y_0)$$
  

$$= x_1 y_1 + (x_1 y_0 + x_0 y_1) + x_0 y_0$$
  

$$(x_1 y_0 + x_0 y_1) = \alpha \beta - x_1 y_1 - x_0 y_0$$

# Key Trick: 4 multiplies at the price of 3

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

$$x = 2^{n/2} \cdot x_1 + x_0 \Rightarrow \alpha = x_1 + x_0$$
  

$$y = 2^{n/2} \cdot y_1 + y_0 \Rightarrow \beta = y_1 + y_0$$
  

$$xy = (2^{n/2} \cdot x_1 + x_0)(2^{n/2} \cdot y_1 + y_0)$$
  

$$= 2^n \cdot x_1 y_1 + 2^{n/2} \cdot (x_1 y_0 + x_0 y_1) + x_0 y_0$$
  

$$A \qquad \alpha \beta - A - B \qquad B$$

To multiply two n-bit integers:

Add two n/2 bit integers.

Multiply three n/2-bit integers.

Add, subtract, and shift n/2-bit integers to obtain result.

$$T(n) = 3T\left(\frac{n}{2}\right) + O(n) \Rightarrow T(n) = O\left(n^{\log_2 3}\right) = O(n^{1.585...})$$

# Integer Multiplication (Summary)

- Naïve:  $\Theta(n^2)$
- Karatsuba:  $\Theta(n^{1.585...})$
- Amusing exercise: generalize Karatsuba to do 5 size n/3 subproblems This gives Θ(n<sup>1.46...</sup>) time algorithm
- Best known algorithm runs in  $\Theta(n \log n)$  using fast Fourier transform

but mostly unused in practice (unless you need really big numbers - a billion digits of  $\pi$ , say)

• Best lower bound O(n): A fundamental open problem

#### Median

## Selecting k-th smallest

Problem: Given numbers  $x_1, ..., x_n$  and an integer  $1 \le k \le n$ output the *k*-th smallest number  $Sel(\{x_1, ..., x_n\}, k)$ 

A simple algorithm: Sort the numbers in time O(n log n) then return the k-th smallest in the array.

Can we do better?

```
Yes, in time O(n) if k = 1 or k = 2.
```

Can we do O(n) for all possible values of k?

Assume all numbers are distinct for simplicity.

# An Idea

Choose a number w from  $x_1, \ldots, x_n$ 

Define

• 
$$S_{<}(w) = \{x_i : x_i < w\}$$

• 
$$S_{=}(w) = \{x_i : x_i = w\}$$

• 
$$S_{>}(w) = \{x_i : x_i > w\}$$

Solve the problem recursively as follows:

- If  $k \leq |S_{\leq}(w)|$ , output  $Sel(S_{\leq}(w), k)$
- Else if  $k \le |S_{\le}(w)| + |S_{=}(w)|$ , output w
- Else output  $Sel(S_{>}(w), k |S_{<}(w)| |S_{=}(w)|)$

Ideally want  $|S_{\leq}(w)|, |S_{\geq}(w)| \le n/2$ . In this case ALG runs in  $O(n) + O\left(\frac{n}{2}\right) + O\left(\frac{n}{4}\right) + \dots + O(1) = O(n)$ .

# How to choose w?

Suppose we choose w uniformly at random similar to the pivot in quicksort.

Then,  $\mathbb{E}[|S_{\leq}(w)|] = \mathbb{E}[|S_{>}(w)|] = n/2$ . Algorithm runs in O(n) in expectation.

Can we get O(n) running time deterministically?

- Partition numbers into sets of size 3.
- Sort each set (takes O(n))
- w = Sel(midpoints, n/6)



# How to lower bound $|S_{<}(w)|, |S_{>}(w)|$ ?



< **w** 

• 
$$|S_{<}(w)| \ge 2\left(\frac{n}{6}\right) = \frac{n}{3}$$
  
•  $|S_{>}(w)| \ge 2\left(\frac{n}{6}\right) = \frac{n}{3}$ .  
 $\frac{n}{3} \le |S_{<}(w)|, |S_{>}(w)| \le \frac{2n}{3}$ 

So, what is the running time?

# Asymptotic Running Time?



- If  $k \leq |S_{\leq}(w)|$ , output  $Sel(S_{\leq}(w), k)$
- Else if  $k \le |S_{\le}(w)| + |S_{=}(w)|$ , output w
- Else output  $Sel(S_>(w), k S_<(w) S_=(w))$

O(nlog n) again? So, what is the point?

Where 
$$\frac{n}{3} \le |S_{\le}(w)|, |S_{>}(w)| \le \frac{2n}{3}$$

$$T(n) = T\left(\frac{n}{3}\right) + T\left(\frac{2n}{3}\right) + O(n) \Rightarrow T(n) = O(n\log n)$$

# An Improved Idea

> w



Partition into n/5 sets. Sort each set and set w = Sel(midpoints, n/10)

•  $|S_{<}(w)| \ge 3\left(\frac{n}{10}\right) = \frac{3n}{10}$ •  $|S_{>}(w)| \ge 3\left(\frac{n}{10}\right) = \frac{3n}{10}$  $T(n) = T\left(\frac{n}{5}\right) + T\left(\frac{7n}{10}\right) + O(n) \Rightarrow T(n) = O(n)$ 

## An Improved Idea

```
Sel(S, k) {
   n \leftarrow |S|
   If (n < ??) return ??
   Partition S into n/5 sets of size 5
   Sort each set of size 5 and let M be the set of medians, so
|M|=n/5
   Let w=Sel(M,n/10)
                                             We can maintain each
   For i=1 to n{
      If x_i < w add x to S_<(w)
                                                 set in an array
      If x_i > w add x to S_>(w)
      If x_i = w add x to S_{=}(w)
   }
   If (k \leq |S_{\leq}(w)|)
      return Sel(S_{\leq}(w), k)
   else if (k \le |S_<(w)| + |S_=(w)|)
      return w;
   else
      return Sel (S_{>}(w), k - |S_{<}(w)| - |S_{=}(w)|)
```

}

# **D&C** Summary

Idea:

"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
- Applications: Many.
  - Binary Search, Merge Sort, (Quicksort),
  - Root of a Function
  - Closest points,
  - Integer multiplication
  - Median
  - Matrix Multiplication