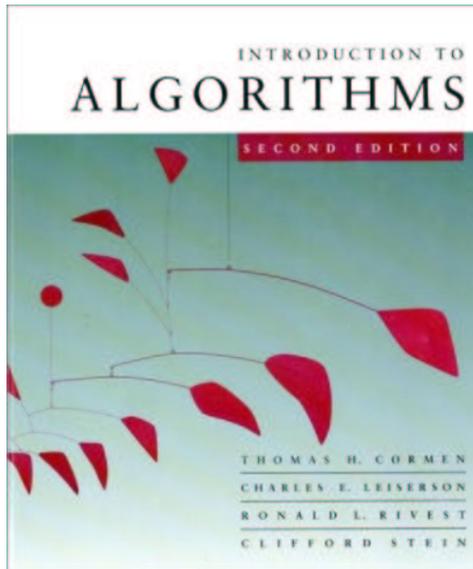


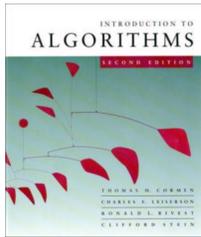
# *Introduction to Algorithms*

## **6.046J/18.401**



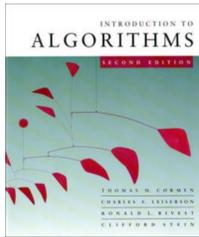
## *Lecture 19*

**Prof. Piotr Indyk**



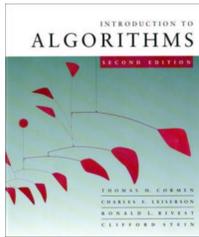
# Computational Geometry

- Algorithms for **geometric** problems
- Applications: CAD, GIS, computer vision,.....
- E.g., the *closest pair* problem:
  - Given: a set of points  $P = \{p_1 \dots p_n\}$  in the plane, such that  $p_i = (x_i, y_i)$
  - Goal: find a pair  $p_i \neq p_j$  that minimizes  $\|p_i - p_j\|$
- We will see more examples in the next lecture



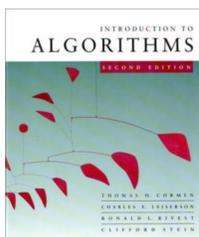
# Computational Model

- In the next two lectures, we will assume that
  - The input (e.g., point coordinates) are *real* numbers
  - We can perform (natural) operations on them in *constant* time, with perfect precision
- Advantage: simplicity
- Drawbacks: highly non-trivial issues:
  - Theoretical: if we allow arbitrary operations on reals, we can compress  $n$  numbers into a one number
  - Practical: algorithm designed for infinite precision sometimes fail on real computers



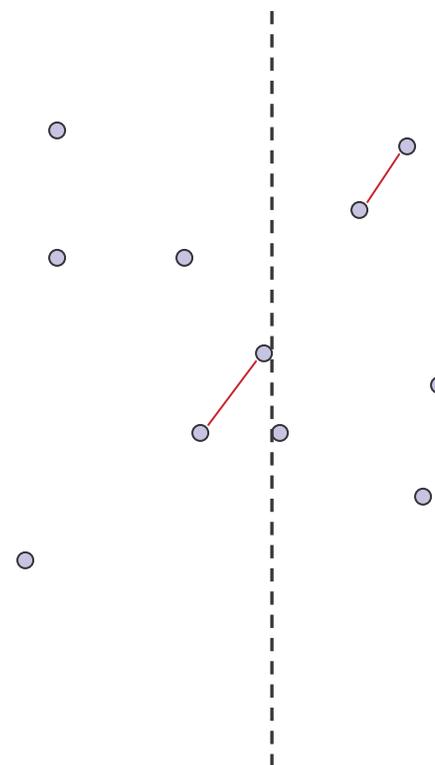
# Closest Pair

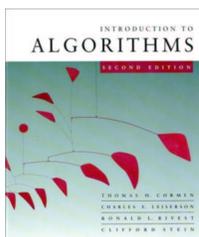
- Find a closest pair among  $p_1 \dots p_n$
- Easy to do in  $O(n^2)$  time
  - For all  $p_i \neq p_j$ , compute  $\|p_i - p_j\|$  and choose the minimum
- We will aim for  $O(n \log n)$  time



# Divide and conquer

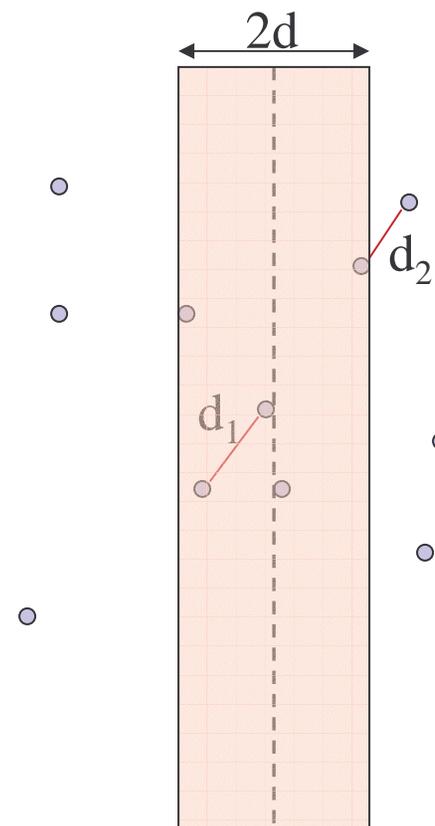
- Divide:
  - Compute the median of x-coordinates
  - Split the points into  $P_L$  and  $P_R$ , each of size  $n/2$
- Conquer: compute the closest pairs for  $P_L$  and  $P_R$
- Combine the results (the hard part)

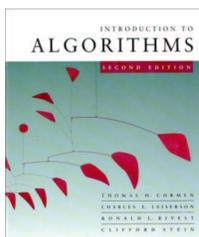




# Combine

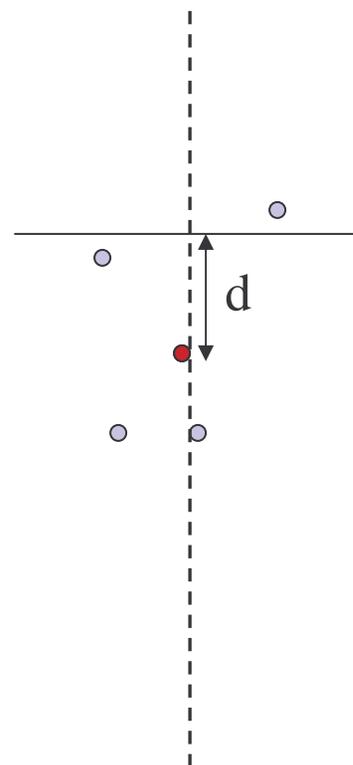
- Let  $d = \min(d_1, d_2)$
- Observe:
  - Need to check only pairs which cross the dividing line
  - Only interested in pairs within distance  $< d$
- Suffices to look at points in the  $2d$ -width strip around the median line

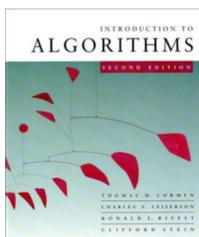




# Scanning the strip

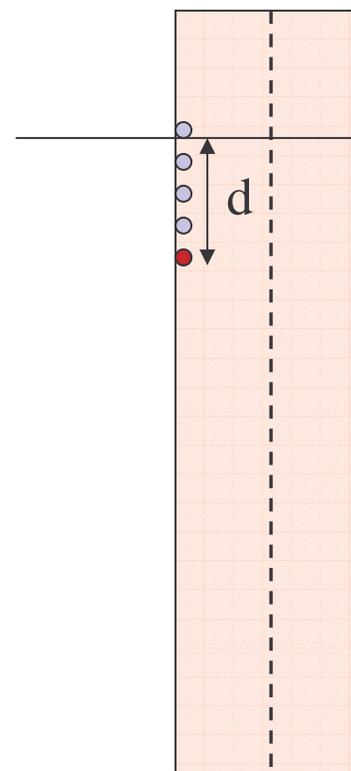
- Sort all points in the strip by their  $y$ -coordinates, forming  $q_1 \dots q_k$ ,  $k \leq n$ .
- Let  $y_i$  be the  $y$ -coordinate of  $q_i$
- For  $i=1$  to  $k$ 
  - $j=i-1$
  - While  $y_i - y_j < d$ 
    - Check the pair  $q_i, q_j$
    - $j:=j-1$

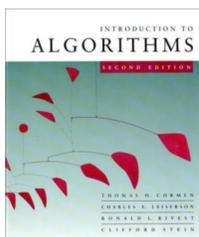




# Analysis

- Correctness: easy
- Running time is more involved
- Can we have many  $q_j$ 's that are within distance  $d$  from  $q_i$  ?
- No
- Proof by *packing* argument



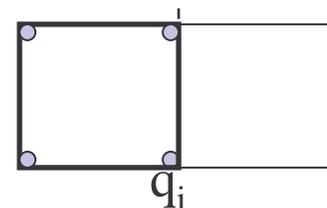


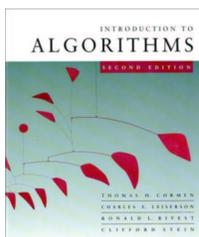
# Analysis, ctd.

**Theorem:** there are at most 7  $q_j$ 's such that  $y_i - y_j \leq d$ .

**Proof:**

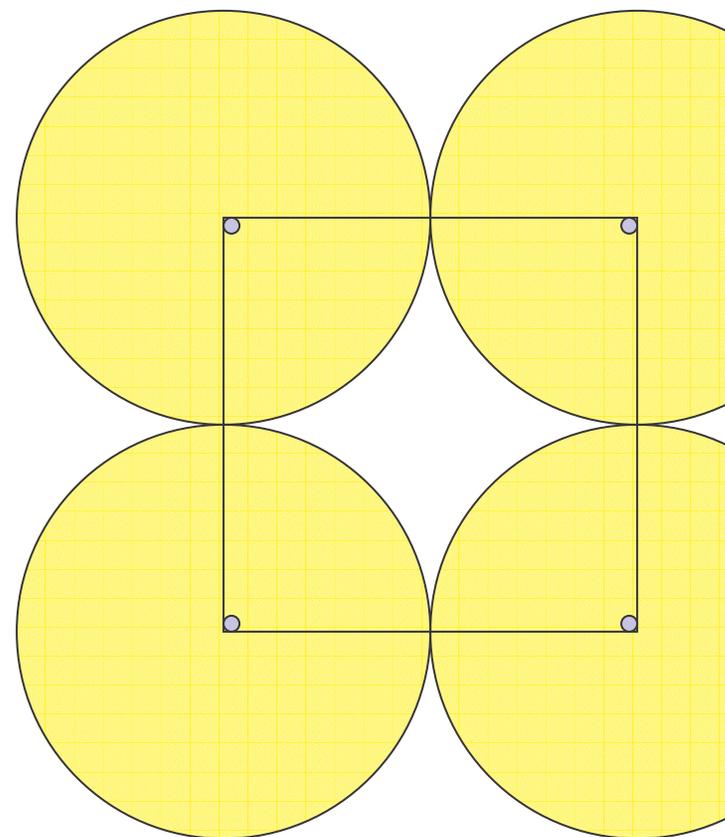
- Each such  $q_j$  must lie either in the left or in the right  $d \times d$  square
- Within each square, all points have distance  $\geq d$  from others
- We can pack at most 4 such points into one square, so we have 8 points total (incl.  $q_i$ )

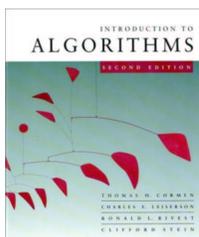




# Packing bound

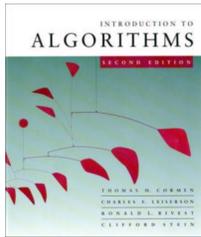
- Proving “4” is not trivial
- Will prove “5”
  - Draw a disk of radius  $d/2$  around each point
  - Disks are disjoint
  - The disk-square intersection has area  $\geq \pi (d/2)^2/4 = \pi/16 d^2$
  - The square has area  $d^2$
  - Can pack at most  $16/\pi \approx 5.1$  points





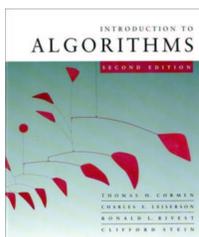
# Running time

- Divide:  $O(n)$
- Combine:  $O(n \log n)$  because we sort by  $y$
- However, we can:
  - Sort all points by  $y$  at the beginning
  - Divide preserves the  $y$ -order of pointsThen combine takes only  $O(n)$
- We get  $T(n)=2T(n/2)+O(n)$ , so  $T(n)=O(n \log n)$



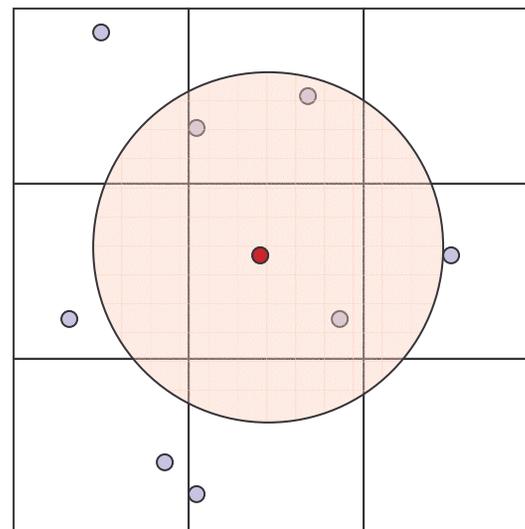
# Close pair

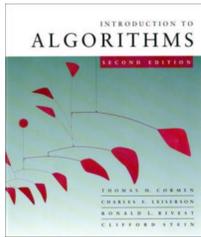
- Given:  $P = \{p_1 \dots p_n\}$
- Goal: check if there is any pair  $p_i \neq p_j$  within distance **1** from each other
- Will give an  $O(n)$  time randomized algorithm, using...  
**... hashing!**



# Algorithm

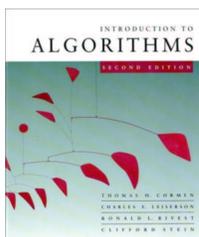
1. Impose a square grid onto the plane, where each cell is a  $1 \times 1$  square
2. Put each point into a bucket corresponding to the cell it belongs to (see last slide)
3. If there is a bucket with  $> 4$  points in it, answer **YES** (by the packing theorem)
4. Otherwise, for each  $p \in P$ , check all points in the cell containing  $p$ , as well as the cells adjacent to it





# Analysis

- Running time:
  - Putting points into the buckets:  $O(n)$  time using hashing
  - Checking if there is a heavy bucket:  $O(n)$
  - Checking the cells:  $9 \times 4 \times n = O(n)$
- Overall: linear time



# To hash or not to hash

- In step 2 of the algorithm, we need to partition the points into “buckets”, i.e., sets  $B_1 \dots B_k$ ,  $k \leq n$ . Each bucket contains all points that belong to some non-empty cell.
- This can be solved using any data structure for the “symbol table” problem, as in Lecture 7. The key of a point  $p=(x,y)$  is the identifier of the cell that  $p$  belongs to. Note that now the keys are not unique, i.e., many points can have the same key.
- We could solve the symbol table problem using direct access table. However, the space used by the algorithm would be proportional to the **total** number of cells in the grid, which could be much larger than  $n$ . In particular, we would not be able to initialize that much space in  $O(n)$  time.
- Hashing allows us to reduce the space (and initialization time) to  $O(n)$ , since the space depends only on the number of **nonempty** cells. Since hashing uses randomness, the resulting algorithm is randomized.