

# CSE520 Computational Geometry

## Lecture 1

### Fixed-Radius Near Neighbors Searching

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# Introduction

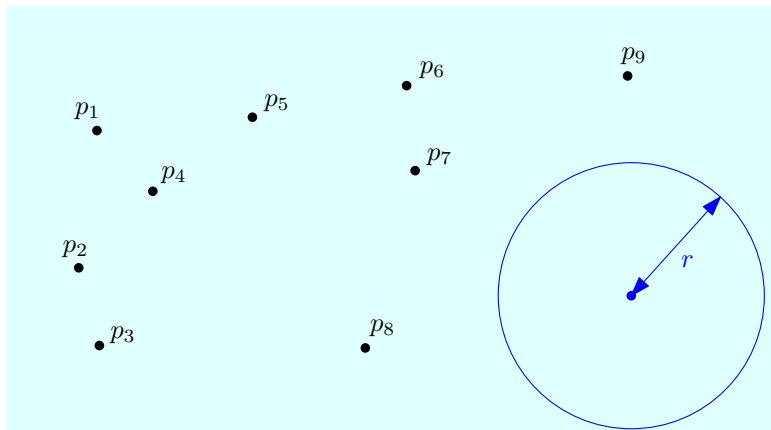
Reference:

- Dave Mount's [lecture notes](#), lecture 2.
- [Wikipedia](#).

An old problem.

- Applications: Air traffic control, molecular dynamics ...
- Simple and very fast algorithm.

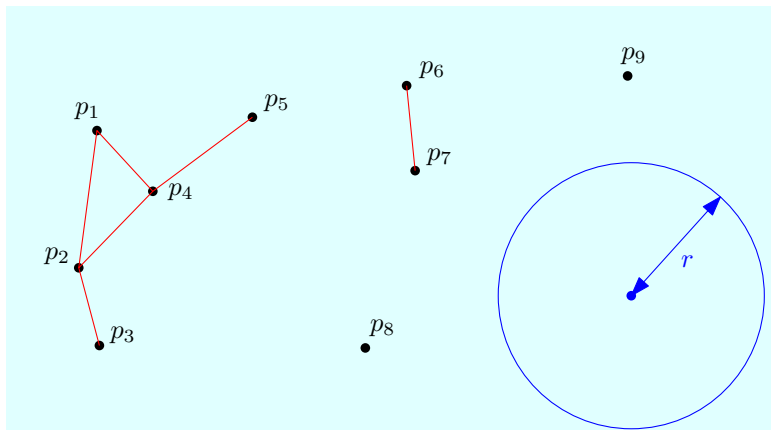
# Problem Formulation



**INPUT:** a set of points  $P = \{p_1, \dots, p_n\}$  and a radius  $r$ .

We assume we are in *fixed dimension*, that is,  $P \subset \mathbb{R}^d$  with  $d = O(1)$ .

# Problem Formulation



**OUTPUT:** all pairs  $(p_i, p_j) \in P^2$  such that the distance  $|p_i p_j|$  is at most  $r$ .

Here, the output is:  $(p_1, p_2), (p_1, p_4), (p_2, p_3), (p_2, p_4), (p_4, p_5), (p_6, p_7)$ .

# Brute-Force Approach

- For any two points  $p_i, p_j$ , we can compute  $|p_i p_j|^2$  in constant time.
  - ▶ Example: if  $d = 2$ ,  $p_i = (x_i, y_i)$  and  $p_j = (x_j, y_j)$ , then

$$|p_i p_j|^2 = (x_i - x_j)^2 + (y_i - y_j)^2.$$

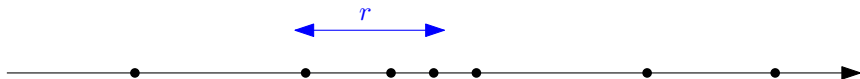
- So we can just check for each pair  $(i, j)$  whether  $|p_i p_j|^2 \leq r^2$ .
  - ▶ Running time?  $\Theta(n^2)$ .
- Can we do better than  $O(n^2)$  ?
  - ▶ Let  $k$  denote the *output size*:  $k$  is the number of pairs  $(p_i, p_j)$  such that  $|p_i p_j| \leq r$ .
  - ▶  $o(n^2)$  is impossible, because in the worst case

$$k = \binom{n}{2} = \frac{n(n-1)}{2} = \Theta(n^2).$$

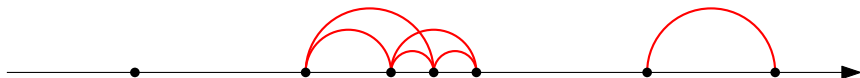
- ▶ So we will give an *output sensitive* algorithm that runs in time  $\Theta(n + k)$ .

# The One-Dimensional Case

- **INPUT:**  $x_1, \dots, x_n \in \mathbb{R}$  and  $r > 0$ .



- **OUTPUT:** the  $k$  pairs  $(x_i, x_j)$  such that  $|x_i - x_j| \leq r$ .



# The One-Dimensional Case

## Exercise

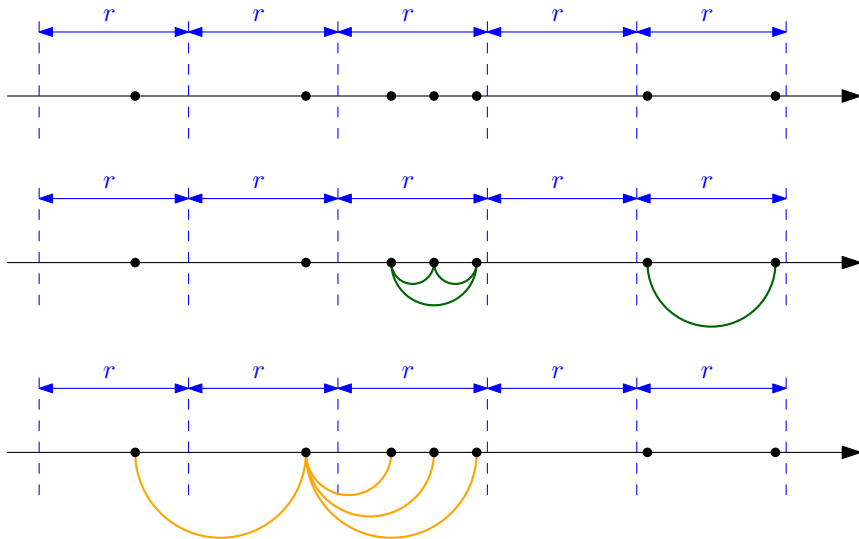
Find a simple  $O(n \log(n) + k)$  algorithm.

## Solution

- Sort the numbers in increasing order.
- Traverse this sorted list from left to right.
- Pseudocode:



# A Bucketing Approach



# A Bucketing Approach

- We partition  $\mathbb{R}$  using intervals (buckets) of size  $r$ :

$$\dots, [-3r, -2r), [-2r, -r), [-r, 0), [0, r), [r, 2r), [2r, 3r) \dots$$

- So each interval  $[br, (b+1)r)$  corresponds to some integer  $b$ .
- A point  $x \in \mathbb{R}$  is in bucket  $b = \lfloor x/r \rfloor$ .
  - ▶ So we can determine the bucket containing  $x$  in  $O(1)$  time.
- Observation: We may only have  $|x_i - x_j| \leq r$  if
  - ▶  $x_i$  and  $x_j$  lie in the same bucket  $b \dots$
  - ▶ or  $x_i$  is in bucket  $b$  and  $x_j$  is in bucket  $b+1$ .
  - ▶ or  $x_j$  is in bucket  $b$  and  $x_i$  is in bucket  $b+1$ .
- So our algorithm only checks pairs in the same bucket, or in adjacent buckets.
- We store the nonempty buckets in a *dictionary* data structure:
  - ▶ It allows deletion, insertion, lookup.

# Pseudocode

## 1D-Bucketing algorithm

- 1: Create an empty dictionary  $D$ .
- 2: **for**  $i \leftarrow 1, n$  **do**
- 3:      $b_i \leftarrow \lfloor x_i / r \rfloor$
- 4:     **if**  $b_i \notin D$  **then**
- 5:         insert bucket  $b_i$  into  $D$
- 6:     insert  $x_i$  into bucket  $b_i$
- 7: **for** each nonempty bucket  $b$  **do**
- 8:     report all pairs  $(x_i, x_j)$  in bucket  $b$
- 9:     report all pairs  $(x_i, x_j) \in b \times (b + 1)$  such that  $x_i - x_j \leq r$

# Analysis

## Theorem

*The bucketing algorithm runs in time  $O(nT(n) + k)$ , where  $T(n)$  is the time needed for one dictionary operation.*

- So it is  $O(n \log(n) + k)$  if we implement the dictionary with a balanced binary search tree,
- and is (randomized) expected time  $O(n + k)$  using a hash table.
- We now prove the theorem.
- First observe that we make  $O(n)$  dictionary operations so they take  $O(nT(n))$  time.
- Line 8 takes time  $O(\sum_{b \in \mathbb{Z}} n_b^2)$  and line 9 takes time  $O(\sum_{b \in \mathbb{Z}} n_b n_{b+1})$  where  $n_b$  is the number of points in bucket  $b$ .
- So Lines 8 and 9 take time  $O(T')$  where

$$T' = \sum_{b \in \mathbb{Z}} n_b^2 + \sum_{b \in \mathbb{Z}} n_b n_{b+1}.$$

# Analysis

- For any  $x, y \in \mathbb{R}$  we have  $xy \leq \frac{x^2 + y^2}{2}$ .
- Therefore

$$\begin{aligned} T' &\leq \sum_{b \in \mathbb{Z}} n_b^2 + \sum_{b \in \mathbb{Z}} \frac{n_b^2 + n_{b+1}^2}{2} \\ &= \frac{3}{2} \sum_{b \in \mathbb{Z}} n_b^2 + \frac{1}{2} \sum_{b \in \mathbb{Z}} n_{b+1}^2 = 2 \sum_{b \in \mathbb{Z}} n_b^2 = 2S. \end{aligned}$$

where  $S = \sum_{b \in \mathbb{Z}} n_b^2$ .

# Analysis

- As all numbers in  $b$  are at distance less than  $r$  from each other, the number of output pairs in bucket  $b$  is

$$k_b = \frac{n_b(n_b - 1)}{2},$$

so

$$S = \sum_{b \in \mathbb{Z}} 2k_b + n_b = 2k + n.$$

- It follows that  $T' \leq 4k + 2n$  hence  $T' = O(n + k)$ .
- Therefore the total running time is  $O(nT(n) + T') = O(nT(n) + n + k) = O(nT(n) + k)$ .

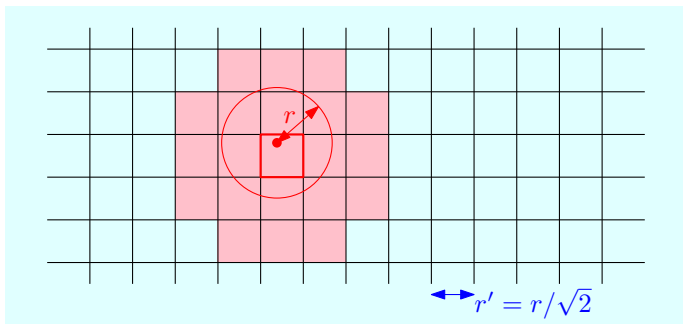
## Two-Dimensional Case

- The bucketing approach applies.
- The integer pair  $(b_x, b_y)$  correspond to the square

$$[b_x r', (b_x + 1)r') \times [b_y r', (b_y + 1)r').$$

where  $r' = r/\sqrt{2}$ .

- For a bucket  $b = (b_x, b_y)$ , we only need to check pairs  $(p, p') \in b \times b'$ , where  $b'$  is one of the 21 buckets pictured below.



# Higher Dimension

- Bucketing applies to any dimension  $d$ .
- When  $d = O(1)$ , it has the same time bound:  $O(nT(n) + k)$ .
- But in general, it is exponential in  $d$ .
- In high dimension, the best known time bounds are not much better than brute force.