

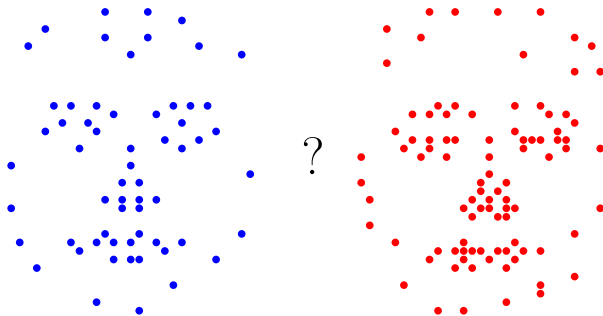
Bipartite Diameter and Other Measures Under Translation

Boris Aronov, **Omrit Filtser**, Matthew J. Katz,
and Khadijeh Sheikhan

September 2019

Similarity between two sets of points

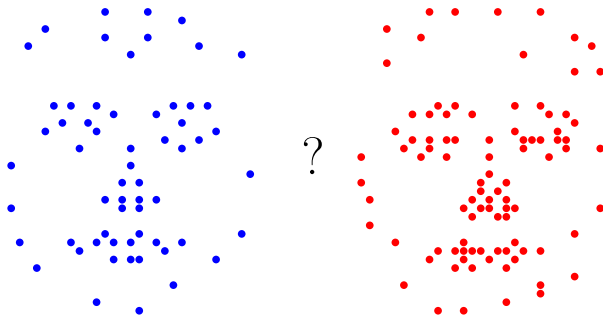
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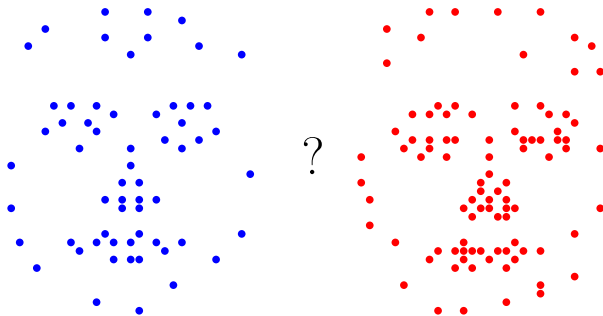
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 - ▶ **Problem:** Sometimes, a bipartite measure is meaningless, unless one of the sets undergoes some transformation.



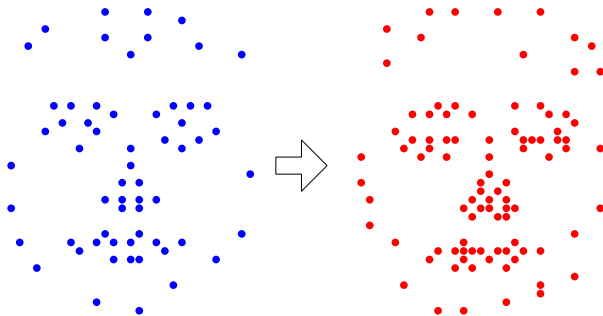
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Find a translation which minimizes some bipartite measure.



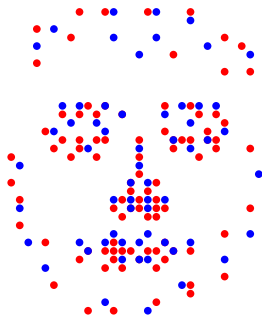
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Bipartite measures under translation

$A = \{a_1, \dots, a_n\}$ and $B = \{b_1, \dots, b_m\}$ – two sets of points in \mathbb{R}^d .

Problem

Find a translation t^ that minimizes **some bipartite measure** of A and $B + t$ over all translations t .*

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Find a translation t^* that minimizes **some bipartite measure** of A and $B + t$ over all translations t .

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- ▶ For the sake of simplicity, we assume that $m = n$.
- ▶ This class of problems naturally extends to other types of transformations, such as rotations, rigid motions, etc.

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- ▶ **Congruence testing**: decide if there exists a transformation that maps A exactly or approximately into B .

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When comparing two sets of points A and B of the same size:

- ▶ **Congruence testing.**
- ▶ **RMS distance:** minimize the sum of squares of distances in a perfect matching between A and B .

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When comparing two sets of points A and B of the same size:

- ▶ **Congruence testing.**
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When comparing two sets of points A and B of different sizes:

- ▶ **Hausdorff distance:** the maximum of the distances from a point in each of the sets to the nearest point in the other set.
Huttenlocher, Kedem, Sharir: $\tilde{O}(n^3)$ in 2D.

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When comparing two sets of points A and B of the same size:

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When comparing two sets of points A and B of different sizes:

- ▶ **Hausdorff distance:** $\tilde{O}(n^3)$ in 2D.
- ▶ Maximum **overlap between the convex hulls** of the sets A and B .
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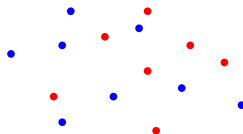
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All the above measures (under various geometric transformations) were widely investigated in the literature.

Our results

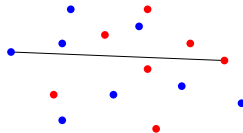
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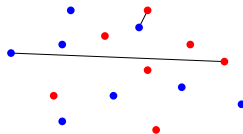
- ▶ **diameter** – the distance between the farthest bichromatic pair, i.e. $\max\{\|a - b\| \mid (a, b) \in A \times B\}$.



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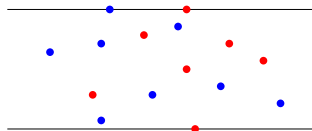
- ▶ **diameter** – $\max\{\|a - b\| \mid (a, b) \in A \times B\}$.
- ▶ **uniformity** – the difference between the bipartite diameter and the distance between the closest bichromatic pair, i.e. $\text{diam}(A, B) - \min\{\|a - b\| \mid (a, b) \in A \times B\}$.



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- ▶ **union width** – the width of $A \cup B$, where the *width* of a set of points in the plane is the smallest distance between a pair of parallel lines, such that the closed strip between the lines contains the entire set.



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Surprisingly, all of these measures (under translation) were not investigated previously in the literature.

Our results

measure	dimension	running time
diameter	$d = 2$	$O(n \log n)$
	$d = 3$	$O(n \log^2 n)$
	$d > 3$ (fixed)	$O(n^2)$
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Diameter

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$$\text{diam}(A, B) = \max\{\|a - b\| \mid (a, b) \in A \times B\}$$

Problem (Bipartite Diameter under Translation)

Find a translation t^ such that for any translation t ,
 $\text{diam}(A, B + t^*) \leq \text{diam}(A, B + t)$.*

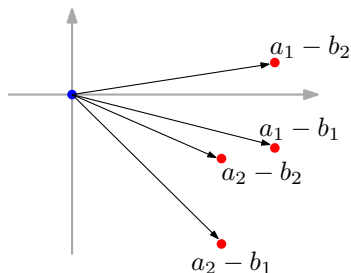
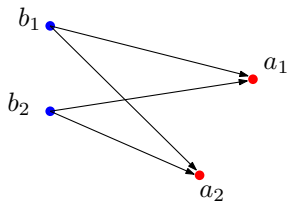
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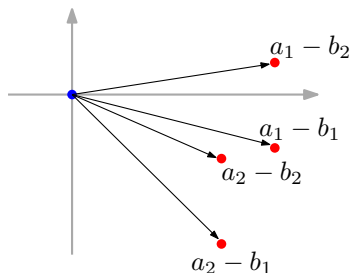
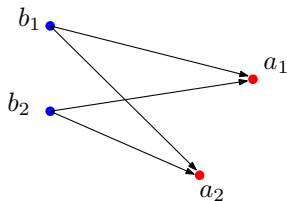
- The set of all possible translations taking a point of B to a point of A .



Diameter

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- ▶ The set of all possible translations taking a point of B to a point of A .
- ▶ Clearly, $|\mathcal{P}| = O(n^2)$.

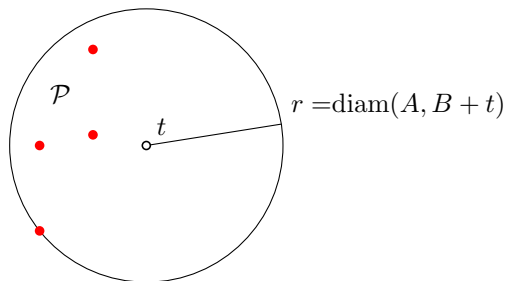


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Claim

Given a point t , the radius of the **minimum enclosing ball** of \mathcal{P} centered at t is equal to $\text{diam}(A, B + t)$.



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Proof.

This radius is

$$\max_{(a-b) \in \mathcal{P}} \|(a-b) - t\| = \max_{(a,b) \in A \times B} \|a - (b+t)\| = \text{diam}(A, B+t).$$



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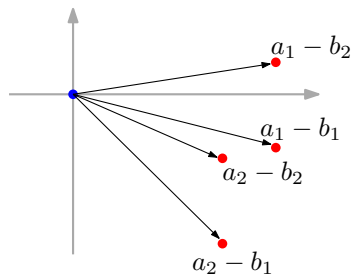
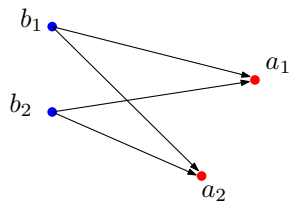
Corollary

The optimal translation t^* minimizing the bipartite diameter coincides with the center of the minimum enclosing ball of \mathcal{P} .

Diameter: Algorithm (naive implementation)

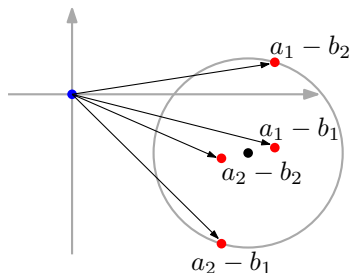
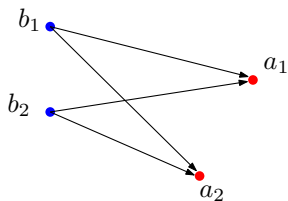
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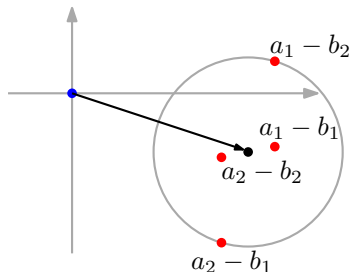
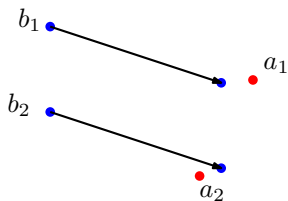
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- ▶ Compute the set of translations \mathcal{P} .
- ▶ Find the center c of the minimum enclosing ball of \mathcal{P} .
- ▶ Translating B by c minimizes the diameter.



Diameter: Running time

The minimum enclosing ball can be computed in:

- ▶ linear time using Megiddo's ('83) algorithm, or
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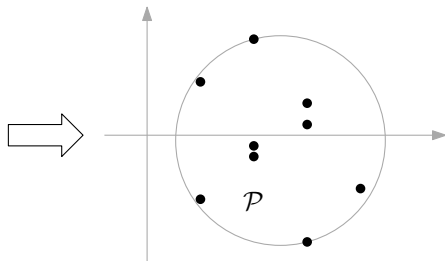
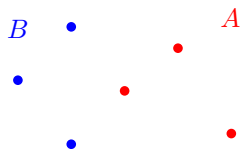
BUT, in 2D and 3D, we can do better!

In fact, computing the minimum enclosing ball of \mathcal{P} in 2D and 3D
(*without computing \mathcal{P} explicitly*) can be done in
near-linear time...

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$$\mathcal{P} = \{a - b \mid (a, b) \in A \times B\}$$

Goal: compute the minimum enclosing ball of \mathcal{P} implicitly.

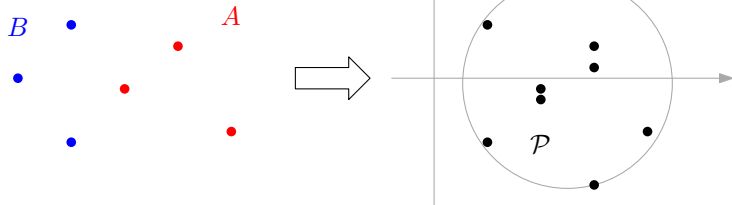


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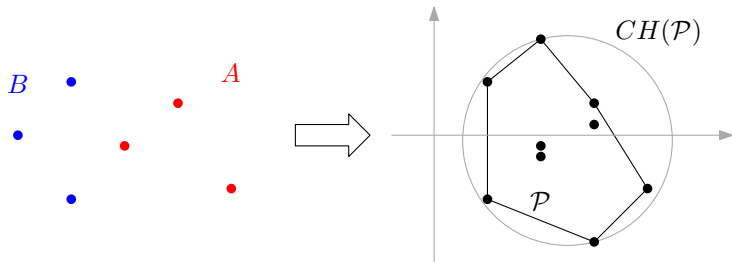
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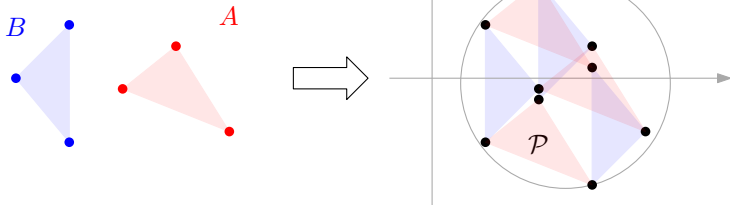
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2. \mathcal{P} is the Minkowski sum of A and $-B$, i.e. $\mathcal{P} = A \oplus -B$.



Diameter in 2D

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A fact from the textbook

For points in 2D, the size of $\text{CH}(A \oplus -B)$ is $O(n)$, and it can be constructed in $O(n)$ time from $\text{CH}(A)$ and $\text{CH}(B)$ using the well-known rotating calipers method...

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$\Rightarrow O(n \log n)$ -time solution for points in 2D!

Diameter in 3D

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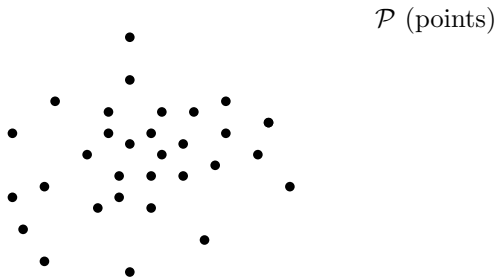
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Idea: The minimum enclosing ball is an LP-type problem \Rightarrow
adapt Clarkson's ('95) scheme for solving LP-type problems.

Diameter in 3D: Algorithm

X – an empty set of points.

Repeat until the minimum enclosing ball is found:

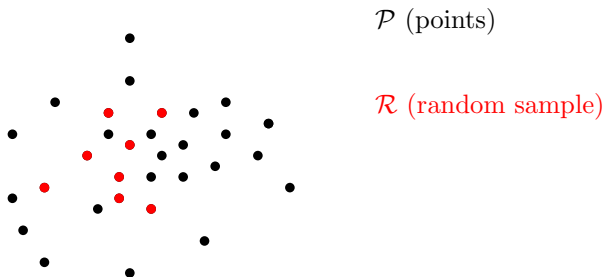


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1: Pick a random sample \mathcal{R} of \mathcal{P} of size $4n$.

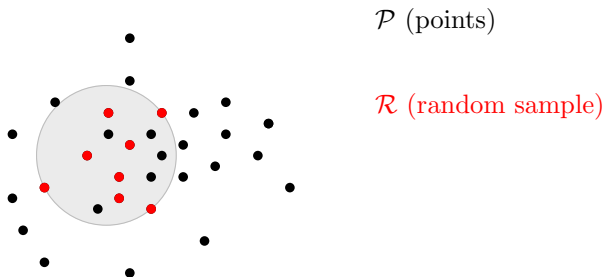


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2: Compute the minimum enclosing ball S of $\mathcal{R} \cup X$.

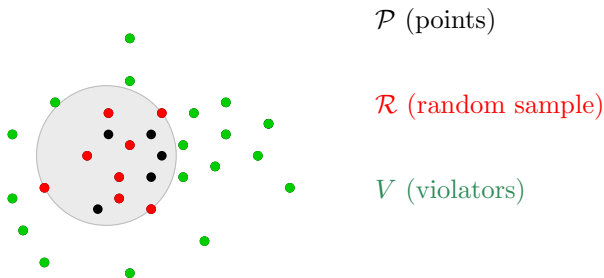


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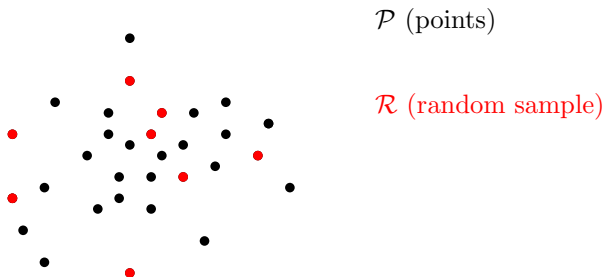
$|V| \geq 2n$, “**bad**” iteration :(

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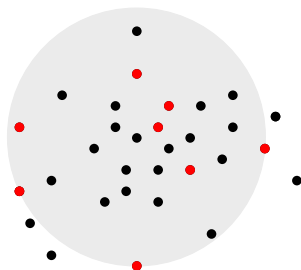


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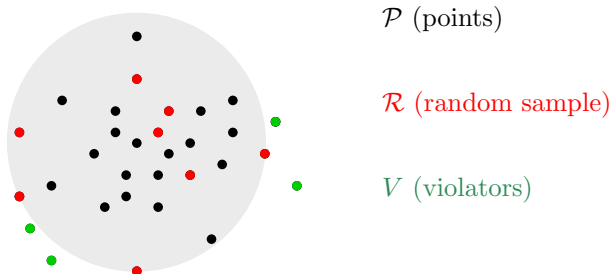
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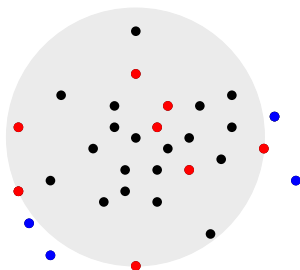
$|V| < 2n$, “good” iteration :)

Diameter in 3D: Algorithm

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4: If $V \neq \emptyset$, then $X \leftarrow X \cup V$ and go to 1. Else, return S .



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V (violators)

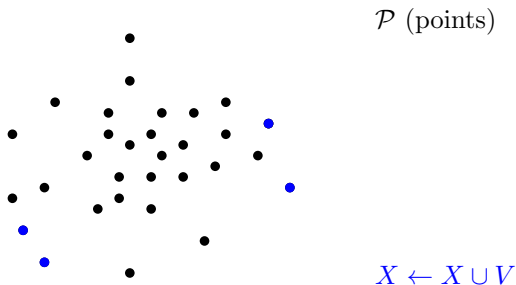
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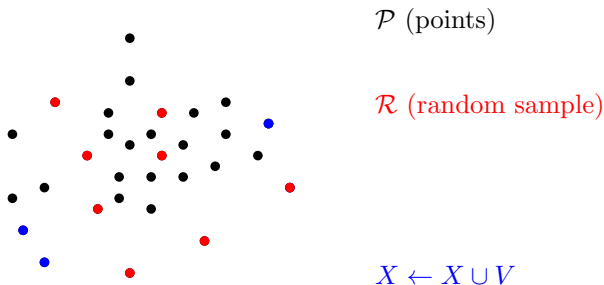


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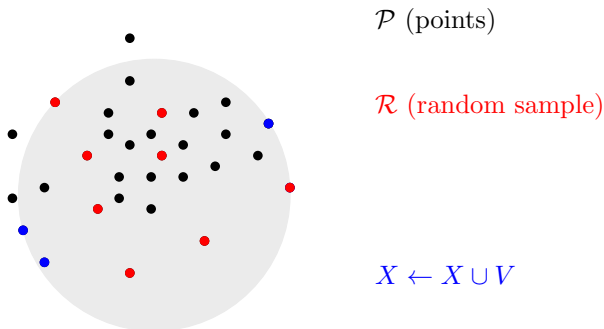


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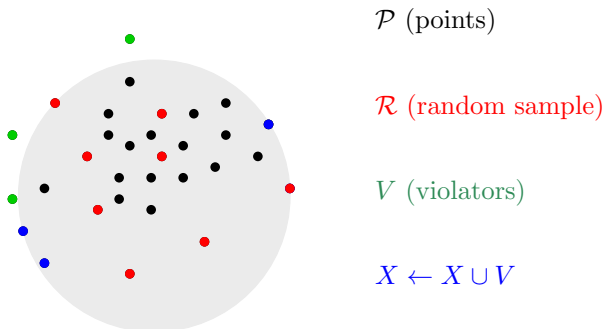


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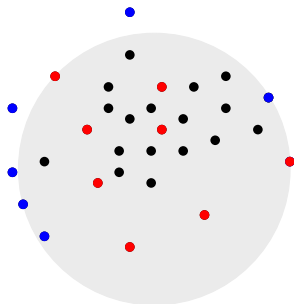


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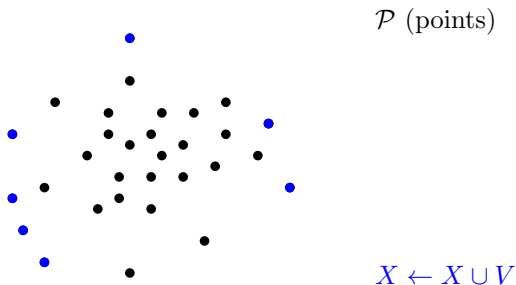
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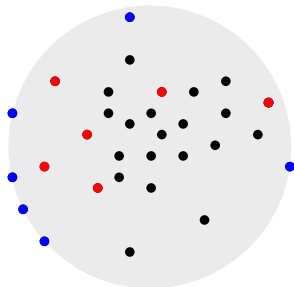


Diameter in 3D: Algorithm

X – an empty set of points.

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\mathcal{P} (points)

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- ▶ Repeatedly pick random points $a \in A$ and $b \in B$ and return $a - b$.

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- ▶ Invoke a standard minimum-ball algorithm on $O(n)$ points, requiring $O(n)$ expected time.

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Given two sets A and B , each of n points in \mathbb{R}^3 , a distance r and a parameter k , report all the pairs of points $a \in A$, $b \in B$ with $\|a - b\| > r$, if there are at most k such pairs. Otherwise, return "TOO MANY".

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$\Rightarrow O(n \log^2 n)$ -time solution for points in 3D!

Uniformity

$A = \{a_1, \dots, a_n\}$ and $B = \{b_1, \dots, b_m\}$ – two sets of points in \mathbb{R}^d .

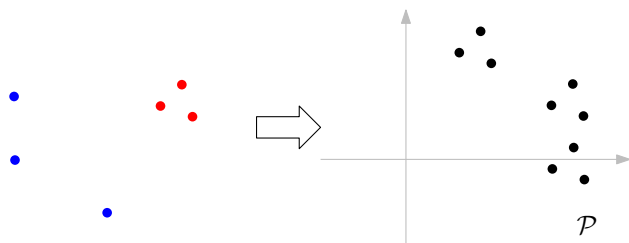
$$\text{uni}(A, B) = \text{diam}(A, B) = \min\{\|a - b\| \mid (a, b) \in A \times B\}$$

Problem (Uniformity under Translation)

Find a translation t^ such that for any translation t ,*
 $\text{uni}(A, B + t^*) \leq \text{uni}(A, B + t)$.

Uniformity

$$\mathcal{P} = \{a - b \mid (a, b) \in A \times B\}$$

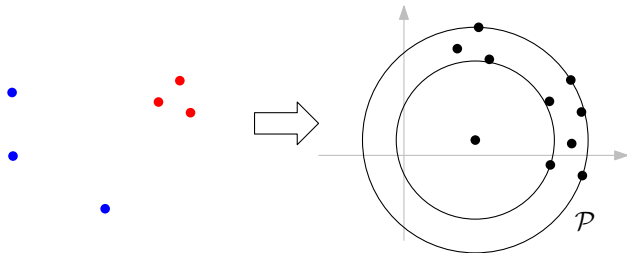


Uniformity

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Claim

The optimal translation t^* minimizing the **uniformity** coincides with the center of the **minimum-width annulus** containing \mathcal{P} .



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- ▶ Agarwal and Sharir ('96): The minimum enclosing annulus of n points in 2D can be computed in $O(n^{3/2+\epsilon})$ expected time...

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The minimum enclosing annulus of n points in 2D
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Thank You!

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 - ▶ Our algorithm (for width in 3D under translation) runs in $O(n^2)$ time. Can we do better?