Bipartite Diameter and Other Measures Under Translation

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



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Goal: Determining the similarity between two sets of points.



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A well investigated problem in computational geometry.





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Find a translation which minimizes some bipartite measure.





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 $A = \{a_1, \ldots, a_n\}$ and $B = \{b_1, \ldots, b_m\}$ – 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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• For the sake of simplicity, we assume that m = n.



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Problem

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

Remarks

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



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

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



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*.



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 the same size:

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When comparing two sets of points A and B of <u>different sizes</u>:

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: Õ(n³) in 2D.



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

- Congruence testing.
- RMS distance.

When comparing two sets of points A and B of <u>different sizes</u>:

- Hausdorff distance: $\tilde{O}(n^3)$ in 2D.
- Maximum overlap between the convex hulls of the sets A and B. de Berg et al.: O(n log n) in 2D, Ahn et al.: Õ(n³) in 3D.



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All the above measures (under various geometric transformations) were widely investigated in the literature.



The main bipartite measures that we consider are:





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► diameter – the distance between the farthest bichromatic pair, i.e. max{||a - b|| | (a, b) ∈ A × B}.





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The main bipartite measures that we consider are:

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





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The main bipartite measures that we consider are:

- diameter max{ $||a b|| | (a, b) \in A \times B$ }.
- uniformity diam(A, B) min $\{ \|a b\| \mid (a, b) \in A \times B \}$.
- ► union width the width of A ∪ 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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- diameter max{ $||a b|| | (a, b) \in A \times B$ }.
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- union width the width of $A \cup B$.
- red-blue width …



The main bipartite measures that we consider are:

- diameter $\max\{||a b|| \mid (a, b) \in A \times B\}$.
- uniformity diam(A, B) min $\{ ||a b|| | (a, b) \in A \times B \}$.
- union width the width of $A \cup B$.
- red-blue width …

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



measure	dimension	running time
diameter	<i>d</i> = 2	$O(n \log n)$
	<i>d</i> = 3	$O(n \log^2 n)$
	d > 3 (fixed)	<i>O</i> (<i>n</i> ²)
uniformity	<i>d</i> = 2	$O(n^{9/4+\varepsilon})$
union width	<i>d</i> = 2	$O(n \log n)$
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 $A = \{a_1, \dots, a_n\} \text{ and } B = \{b_1, \dots, b_m\} - \text{two sets of points in } \mathbb{R}^d.$ $diam(A, B) = \max\{||a - b|| \mid (a, b) \in A \times B\}$

Problem (Bipartite Diameter under Translation)

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



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



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

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



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

• Clearly,
$$|\mathcal{P}| = O(n^2)$$
.



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

Claim

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





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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 diam(A, B + t).

Proof.

This radius is at most

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



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

Claim

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

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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Compute the set of translations *P*.



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- Compute the set of translations *P*.
- Find the center c of the minimum enclosing ball of \mathcal{P} .


Diameter: Algorithm (naive implementation)

- Compute the set of translations *P*.
- ▶ Find the center *c* of the minimum enclosing ball of *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
- expected linear time using Welzl's ('91) simpler randomized algorithm.



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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...



 $\mathcal{P} = \{a - b \mid (a, b) \in A \times B\}$ **Goal**: compute the minimum enclosing ball of \mathcal{P} implicitly.



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

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



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

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

Observations

1. We only need to look at the **convex hull** (CH) of \mathcal{P} .



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

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

Observations

- 1. We only need to look at the **convex hull** (CH) of \mathcal{P} .
- 2. \mathcal{P} is the Minkowski sum of A and -B, i.e. $\mathcal{P} = A \oplus -B$.



Diameter in 2D

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

Goal: compute the minimum enclosing ball of \mathcal{P} in 2D.



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Diameter in 2D

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Goal: compute the minimum enclosing ball of \mathcal{P} in 2D.

A fact from the textbook

For points in 2D, the size of $CH(A \oplus -B)$ is O(n), and it can be constructed in O(n) time from CH(A) and 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

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

Goal: compute the minimum enclosing ball of \mathcal{P} in 3D.



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Diameter in 3D

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

Goal: compute the minimum enclosing ball of \mathcal{P} in 3D.

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



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





```
\mathcal{R} (random sample)
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2: Compute the minimum enclosing ball S of $\mathcal{R} \cup X$.



 \mathcal{P} (points)

 \mathcal{R} (random sample)



X – an empty set of points. Repeat until the minimum enclosing ball is found:

3: Find the set of *violators* V. If $|V| \ge 2n$, go to 1.





 \mathcal{R} (random sample)

V (violators)



$|V| \ge 2n$, "bad" iteration :(

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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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 $X \leftarrow X \cup V$



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- \Rightarrow Expected number of iterations is constant!



Diameter in 3D: implementing an iteration

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What is the running time for one iteration?

- 1: Pick a random sample \mathcal{R} of \mathcal{P} of size 4n.
- ▶ Repeatedly pick random points a ∈ A and b ∈ B and return a − b.



What is the running time for one iteration?

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What is the running time for one iteration?

- 2: Compute the minimum enclosing ball S of $\mathcal{R} \cup X$.
- Invoke a standard minimum-ball algorithm on O(n) points, requiring O(n) expected time.



What is the running time for one iteration?

3: Find the set of violators V.



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▶ ?



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Idea: We solve the following problem.

Problem

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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• Expected running time $O((n+k)\log^2 n)$.



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• Expected running time $O((n + k) \log^2 n)$.

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



Uniformity

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

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

Problem (Uniformity under Translation)

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



Uniformity

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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+ε}) expected time...



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Claim

The minimum enclosing annulus of n points in 2D — with only $O(\sqrt{n})$ extreme points can be computed in $O(n^{9/8+\varepsilon})$ expected time.



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The minimum enclosing annulus of n points in 2D — with only $O(\sqrt{n})$ extreme points can be computed in $O(n^{9/8+\varepsilon})$ expected time.

 $\Rightarrow O(n^{9/4+\varepsilon})$ -time solution!



Thank You!



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Consider other types of transformations?



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 - For width in 3D (without translation) there is an $O(n^{3/2+\epsilon})$ -time algorithm (Agarwal and Sharir).



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 - Consider minimum ratio instead of minimum difference?
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- Width:
 - For width in 3D (without translation) there is an O(n^{3/2+ϵ})-time algorithm (Agarwal and Sharir).
 - Our algorithm (for width in 3D under translation) runs in O(n²) time. Can we do better?

