Approximate Line Segment Nearest Neighbor Search amid Polyhedra in 3-Space

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Problem statement

Input:

A set Π of **polyhedra** with total complexity *n* in \mathbb{R}^3 .

Objective:

Preprocess Π so that given any **query line segment** *s*, one may quickly find the **polyhedron in** Π **nearest to** *s*. In this study, we consider the **approximate** version:

For any real parameter ε > 0, return an input polyhedron whose distance to s is at most (1 + ε) times the distance from the nearest input polyhedron to s.

Motivation:

Nearest neighbor search has applications in areas such as computational geometry, machine learning, data science, etc.

A particular relevance in **path planning** problems involving **collision or clearance queries** for **non-point objects** in three-dimensional space.

How **difficult** is the problem? To get a rough idea...

NNS with *O*(log *n*) query time has long been closely connected to Voronoi diagram (VD).

- In R^d, the worst-case complexity of VD for **point sites** is Θ(n^[d/2]).
 Note: Nearest neighbor search for non-point objects is at least as hard as for point objects.
- In \mathbb{R}^3 , the worst-case complexity of VD for general sites is not well understood.
 - For a few non-point sites and metrics in R³, known lower bounds are roughly quadratic (e.g., see [Sharir, 1995], [Boissonnat et al., 1998], [Har-Peled, 2001], [Koltun and Sharir, 2004]).
- For any "reasonable" class of geometric sites, the maximum complexity of VD in R^d is conjectured to be close to Θ(n^{d-1}) [Boris, 2001].

Both VD- and non-VD-style data structures have been proposed for NNS problems.

Input and query objects are **points**:

- Efficient exact algorithms for low dimensions (e.g., see [Clarkson, 1988], [Meiser, 1993]).
- Efficient **approximate** algorithms for **high dimensions** (e.g., see [Andoni et al., 2014], [Andoni and Indyk, 2017], [Arya et al., 2017]).

Input objects	Query object	References
Polyhedra in \mathbb{R}^3	- - Point -	Koltun and Sharir, 2004
Lines		Chew et al., 1998; Har-Peled, 2001; Mahabadi, 2014
k-flats		Magen, 2007; Basri et al., 2010; Agarwal et al., 2017
Line segments		Abdelkader and Mount, 2021
Points	Line	Andoni et al., 2009
	<i>k</i> -flat	Mulzer et al., 2015
	Line segment in \mathbb{R}^2	Bespamyatnikh, 2003; Goswami et al., 2004; Segal and Zeitlin, 2008
Polygons in \mathbb{R}^2	Line segment	Daescu and Malik, 2018

Input and/or query objects are **more complex than points**:

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Polygons in \mathbb{R}^2	Line segment	Daescu and Malik, 2018

Input and/or query objects are more complex than points:

- Most recent (related) work [Abdelkader and Mount, 2021]:
 - Approximate nearest neighboring line segment to a query point in \mathbb{R}^d .
 - $O((n^2/\varepsilon^d) \log (\Delta/\varepsilon))$ preprocessing time/space, $O(\log(\max\{n, \Delta\}/\varepsilon))$ query time, where Δ is the spread of the input line segments.

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Polygons in \mathbb{R}^2	Line segment	Daescu and Malik, 2018

Input and/or query objects are more complex than points:

- Two-dimensional variant of our problem [Daescu and Malik, 2018]:
 - Exact nearest neighboring polygon to a query line segment in the plane.
 - $O(m \log n)$ preprocessing time, O(m) space, $O((n/m^{1/2}) \operatorname{polylog} n)$ query time, for any $n \le m \le n^2$.

Overview of our results

A non-trivial algorithm for $(1 + \varepsilon)$ -approximate line segment nearest neighbor search among polyhedra in \mathbb{R}^3 .

For any $n \le m \le n^3$, we obtain

 $O((m/\varepsilon) \operatorname{polylog} n + n^{2+\varepsilon})$ preprocessing time/space and $O((1/\varepsilon)(n/m^{1/3}) \operatorname{polylog} n)$ query time.

Specifically, if we set m = n, we have

 $O(n^{2+\varepsilon})$ preprocessing time/space and $O((n^{2/3}/\varepsilon)$ polylog n) query time.



Machineries and tools used:

Polyhedral metric (for approximating the Euclidean distance), approximate Voronoi diagram, real algebraic geometry, multi-level partition trees.

Approximate line segment nearest neighbor

Suppose that query line segment *s* intersects a polyhedron in Π .

• Can be identified in $O(n^{1/2+\alpha})$ time after a preprocessing that takes $O(n^{3/2+\alpha})$ expected time and space [Ezra and Sharir, 2022].

Hereafter, assume that *s* **does not intersect** any polyhedron in Π .

- 1) Find the polyhedron in Π closest to each endpoint of *s* (*a* and *b*).
- 2) Find the polyhedron in $\Pi \cap S_{ab}$ closest to *s* (i.e., closest orthogonal neighbor to *s*).

Of the polyhedra found above, the one with the shortest distance to s is the polyhedron in Π nearest to s.



Nearest neighbor to each endpoint of *s*

Subproblem 1.

Given a set Π of polyhedra with total complexity *n* in \mathbb{R}^3 , preprocess Π so that for any query point *p*, one can quickly determine the polyhedron in Π closest to *p*.

Exact solution:

- Let T be the set of O(n) triangular faces of the polyhedral in Π .
- Reduce Subproblem 1 to finding the triangle of *T* nearest to *p*.
- For each triangle $\tau \in T$, define $f_{\tau}(p)$ to be the Euclidean distance from any point p to τ .
- Let $C_T = \{ f_\tau(p) \mid \tau \in T \}.$
- Let M_T be the **lower envelope** of C_T .
- For any query point p = (x, y, z), the triangle $\tau \in T$ nearest to p is given by $f_{\tau}(p)$ attaining M_T at (x, y, z).
- For any α > 0, lower envelope M_T can be constructed in O(n^{3+α}) expected time and stored in a data structure of O(n^{3+α}) size such that a nearest triangle search query can be answered in O(log² n) time [Agarwal et al., 1997].

Nearest neighbor to each endpoint of *s*

Subproblem 1.

Given a set Π of polyhedra with total complexity *n* in \mathbb{R}^3 , preprocess Π so that for any query point *p*, one can quickly determine the polyhedron in Π closest to *p*.

Approximate solution:

• Define the **polyhedral distance** between any two points *p* and *q*:

 $d_Q(p, q) = \sup\{t \mid q \notin p + tQ\},\$

where Q is a symmetric convex polytope.

 For any ε > 0, there is a Q represented by the intersection of O(1/ε) half-spaces such that
 d(p, q) ≤ d_Q(p, q) ≤ (1 + ε) d(p, q),

where d(p, q) is the Euclidean distance between p and q [Dudley, 1974].

- Construct the Voronoi diagram V of Π under d_Q in O(n^{2+ε}) time, and the complexity of V is O(n^{2+ε}) [Kolton and Sharir, 2004].
- Using V, for any query point p, we can report, in O(log n) time, a (1 + ε)-approximate nearest neighboring polyhedron in Π to p.

Nearest orthogonal neighbor to s

Subproblem 2.

Given a set Π of polyhedra with total complexity *n* in \mathbb{R}^3 , preprocess Π so that for any query line segment *s*, one can quickly determine the polyhedron in $\Pi \cap S_{ab}$ closest to *s*.

Approximate using a **polyhedral metric**:

• Define a convex polygonal prism *T* that is axially symmetrical to *s*: s + tQ', where Q' is a symmetric convex $O(1/\varepsilon)$ -gon.

T has two $O(1/\varepsilon)$ -gons Q'_a and Q'_b as base faces, connected by $O(1/\varepsilon)$ rectangular sides.

• Using *T*, define the polyhedral distance between *s* and a polyhedron *P* in $\Pi \cap S_{ab}$:

 $d_{Q'}(p, q) = \sup\{t \mid P \cap (s + tQ') = \emptyset\}.$



Process each of $O(1/\varepsilon)$ faces and edges of *T* for face- and edge-shooting queries.

Two scenarios to be considered: A) A shooting face hits a vertex of *P*.

B) A **shooting edge** hits an **edge** of *P*.

Scenario A

A query shoots a **fixed-direction rectangular face** from *s*.

The expanding face traces a **3D infinite wedge**.

Look for the first time the expanding wedge hits a vertex of an input polyhedron.



Let V be the set of O(n) vertices of the polyhedra in Π .

Construct a **5-level partition tree** on *V*:

• First 4 levels are used to collect the vertices in *V* that lie within the infinite wedge as the union of a small number of canonical subsets.

Each of first 4 levels supports half-space range searching queries among the points of V.

- 5-th level supports queries that ask for the vertex in the canonical subsets that is minimal in a fixed direction.
- Following standard methodology for constructing multi-level partition trees [Agarwal, 2017; Chan, 2012; Dobkin and Edelsbrunner, 1987; Matoušek, 1993]:

Preprocessing time/space is O(m polylog n), and query time is $O(n/m^{1/3} \text{ polylog } n)$, for any $n \le m \le n^3$.

Scenario B

Two types of shooting edges:





(I) Normal to *s*

(II) Parallel to s

Type I

A query shoots a **fixed line segment normal to** *s* from an endpoint of *s*.

The expanding line segment traces a **2D infinite wedge**.

Seek for the first time the expanding wedge hits an edge of an input polyhedron.

This scenario has been taken care of when solving Subproblem 1.

Scenario B – Type II

A query shoots a **fixed line segment** identical and parallel to *s* from *s*.

The moving line segment traces an **infinite rectangle**.

Look for the first time the expanding rectangle hits an edge of an input polyhedron.



Let *E* be the set of O(n) edges of the polyhedra in Π . Let *e* denote the fixed shooting edge parallel to *s*. Assume that *e* and *s* are in some plane $z = z_0$.

Let ℓ be the length of *s*.

Parameterize each edge η in *E*:

$$x = u_x(\eta)z + v_x(\eta)$$
$$y = u_y(\eta)z + v_y(\eta)$$
$$z = u_z(\eta)z + v_z(\eta)$$

where $u_x(\eta)$, $v_x(\eta)$, $u_y(\eta)$, $v_y(\eta)$, $u_z(\eta)$, $v_z(\eta) \in \mathbb{R}$.



Scenario B – Type II



• First 4 levels are to collect the edges in *E* that satisfies (1).

Define 4 planar point sets $P_1 = \{(u_y(\eta), v_y(\eta)) \mid \eta \in E\}, P_2 = \{(1, -u_z(\eta)) \mid \eta \in E\}, P_3 = \{(1, -v_z(\eta)) \mid \eta \in E\}, \text{ and } P_4 = \{(u_x(\eta), v_x(\eta)) \mid \eta \in E\}.$

1-st and 2-nd levels support half-plane range searching queries against P_1 , 3-rd level against P_2 , and 4-th level against P_3 .

- 5-th level supports queries that ask for the edge that is minimal in a fixed direction.
- Preprocessing time/space is O(m polylog n), and query time is $O(n/m^{1/2} \text{ polylog } n)$, for any $n \le m \le n^2$.

Nearest orthogonal neighbor to s

Subproblem 2.

Given a set Π of polyhedra with total complexity *n* in \mathbb{R}^3 , preprocess Π so that for any query line segment *s*, one can quickly determine the polyhedron in $\Pi \cap S_{ab}$ closest to *s*.

Prepare query data structures:

• For each shooting face of *T* in Scenario A:

O(m polylog n) preprocessing time/space, $O(n/m^{1/3} \text{ polylog } n)$ query time, for any $n \le m \le n^3$.

• For each shooting edge of *T* in Scenario B – Type II:

O(m polylog n) preprocessing time/space, $O(n/m^{1/2} \text{ polylog } n)$ query time, for any $n \le m \le n^2$.

Since there are $O(1/\varepsilon)$ such faces and edges, this takes $O((m/\varepsilon) \operatorname{polylog} n)$ preprocessing time and storage total, for any $n \le m \le n^3$.

Total cost of a query is $O((1/\varepsilon)(n/m^{1/3}) \text{ polylog } n)$.



Overview of our results

Subproblem 1: Finding nearest neighbor to each endpoint of query line segment *s*

 $O(n^{2+\varepsilon})$ preprocessing time/space, $O(\log n)$ query time.

Subproblem 2: Finding nearest orthogonal neighbor to query line segment *s*

 $O((m/\varepsilon) \text{ polylog } n)$ preprocessing time/space, $O((1/\varepsilon)(n/m^{1/3}) \text{ polylog } n)$ query time, for any $n \le m \le n^3$.

In conclusion, for any $n \le m \le n^3$, we can find a $(1 + \varepsilon)$ -approximate line segment nearest neighbor among polyhedra in \mathbb{R}^3 with

 $O((m/\varepsilon) \text{ polylog } n + n^{2+\varepsilon})$ preprocessing time/space and $O((1/\varepsilon)(n/m^{1/3}) \text{ polylog } n)$ query time.

Specifically,

m = n $O(n^{2+\varepsilon})$ preprocessing time/space and $O((n^{2/3}/\varepsilon) \operatorname{polylog} n)$ query time.

 $m = n^3$ $O((n^3/\varepsilon) \text{ polylog } n)$ preprocessing time/space and $O((1/\varepsilon) \text{ polylog } n)$ query time.

Concluding remarks

Partly motivated by **path planning applications**, where **paths** are suggested in real time and need to be verified quickly if they **satisfy certain constraints** such as having a given clearance from obstacles.

Given a set of polyhedra in 3-space, preprocess them so that for any query polygonal path and any real value c > 0, one can quickly

i) report the clearance of the path, and/or

ii) determine if the path has a clearance of at least *c*.

Query (i) can be answered by finding the exact nearest neighboring input polyhedron.

Query (ii) can be addressed after performing query (i).

Unfortunately, exact nearest neighbor search in such a setting is expensive, and approximation (using the results herein) may yield inconclusive answer.

Is it feasible to quickly answer query (ii) definitively without an exact solution to query (i)?

Note: After obtaining a decision algorithm for query (ii), query (i) can be answered using parametric search.

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Thank you!

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