An Efficient Algorithm for Determining the Connected Orthogonal Convex hull

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Abstract The Quickhull algorithm for finding the convex hull of a finite set of points was independently conducted by Eddy in 1977 and Bykat in 1978. Inspired by the idea of this algorithm, we present a new efficient algorithm for finding extreme points of the connected orthogonal convex hull of a finite set of points that still keeps advantages of the quickhull algorithm. Consequently, our algorithm runs faster than the others (the algorithms introduced by Montuno and Fournier in 1982 and by An, Huyen and Le in 2020). We also show that the expected complexity of the algorithm is $O(n \log n)$, where n is the number of points.

Keywords Convexity · Extreme points · Quickhull algorithm · Orthogonal convex hulls · x-y convex hulls.

Mathematics Subject Classification (2000) MSC $52A30 \cdot MSC$ $52B55 \cdot MSC$ $68Q25 \cdot MSC$ 65D18.

1 Introduction

Orthogonal convexity (rectilinearity, or (x,y) convexity, or x-y convexity) is one of the most extensively subjects studied in computational geometry and convex analysis. It is widely used in research fields, including illumination [1], polyhedron reconstruction [9], geometric search [16], and VLSI circuit layout

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design [17], digital images processing [15]. The concept of orthogonal convex hull of a set was first mentioned in 1959 by Unger [18]. By 1983, Montuno and Fournier introduced efficient algorithms for computing the (x, y)-convex hulls of a finite set of planar points, an (x, y)-polygon and of a set of (x, y)-polygons under various conditions [11]. The condition under which the (x, y)-convex hull exists is given and an algorithm for testing if the given set of (x, y)-polygons satisfies the condition is also presented. Since then, there exist several algorithms for finding orthogonal convex hull proposed [10,12], and [14]. Recently, in 2020, An, Huyen and Le give a condition that ensures the unique of the orthogonal convex hull of a finite planar point set and determine the hull through its extreme points [3]. Their efficient algorithm is modified from the Graham's convex hull algorithm. They also show that the lower bound of computational complexity of such algorithms is $O(n \log n)$.

The Quickhull algorithm [4,6,7,13] determines the convex hull of a finite planar set of points. The worst case complexity of the algorithm is $O(n^2)$ and its average time is $O(n \log n)$. Quickhull is known as a powerful algorithm, which runs in practice much faster than in the worst case. The recursive nature of the Quickhull algorithm allows a fast implementation. This algorithm can also be easily designed as a parallel algorithm for finding convex hull of the point set. Recognizing the effectiveness of the Quickhull algorithm, in this paper, we apply the idea of this algorithm and its improved algorithm [8] to propose an algorithm, namely \mathcal{O} -Quickhull, for determining extreme points of the orthogonal convex hull of a finite set of points and compare it with the algorithm [3] and Montuno and Fournier's algorithm [11]. We also show that the expected complexity of \mathcal{O} -Quickhull is $O(n \log n)$, where n is the number of points.

The paper consists of several sections. Section 2 presents some concepts of connected orthogonal convexity that will be used in this paper. Section 3 introduces the definition of directed orthogonal lines and some other concepts. Section 4 is devoted to an algorithm based on the Quickhull algorithm to determine extreme points of the connected orthogonal convex hull of a finite planar point set and its expected complexity. Section 5 closes the paper with some numerical experiments.

2 Connected orthogonal convex hulls and their properties

Throughout this paper, we focus on the problem of determining the *connected* orthogonal convex hull of a finite planar point set.

Let be given $p, q, t \in \mathbb{R}^2$, denote $[p, q] := \{(1 - \lambda)p + \lambda q : 0 \le \lambda \le 1\}$, pq the straight line through the points p and q and $\mathrm{dist}(t, pq)$ the Euclidean distance from t to the line pq. We denote by x_p and y_p respectively the x-coordinate and y-coordinate of p. As usual, $\mathrm{dist}(p, q) := \sqrt{(x_p - x_q)^2 + (y_p - y_p)^2}$.

Definition 1 ([18]) A set $S \subset \mathbb{R}^2$ is said to be *orthogonal convex* if its intersection with any horizontal or vertical line is convex.

 ${\cal S}$ is said to be ${\it connected\ orthogonal\ convex}$ if it is orthogonal convex and connected.

Definition 2 ([14]) A connected orthogonal convex hull of S is a smallest connected orthogonal convex set containing S.

We define a line to be *rectilinear* if the line is parallel to either x-axis or y-axis. A half line or a line segment are *rectilinear* if the lines on which they lie are rectilinear.

Let $p \neq q$ be two given points in the plane. We define $l(p,q)(x_p \neq x_q, y_p \neq y_q)$ through p,q to be union of two rectilinear half lines having the same starting point. If $x_p = x_q$ or $y_p = y_q$ then l(p,q) is the line through p and q. The set l(p,q) is called the *orthogonal line* through p and q. The common point of two the rectilinear half lines of l(p,q) is called the *vertex* of l(p,q). We also denote by $l^v(p,q)$ the orthogonal line l(p,q) having the vertex v.

An orthogonal line $l(p,q), (x_p \neq x_q, y_p \neq y_q)$ separates the plane into two regions. The quadrant region together with the orthogonal line l(p,q) will be called a quadrant determined by the orthogonal line.

Definition 3 ([3]) Given a set $S \subset \mathbb{R}^2$. An l(p,q) is an orthogonal supporting line (O-support, for brevity) of a set S (p and q might not belong to S) if the intersection of l(p,q) with S is non-empty and either all points of $S \setminus (S \cap l(p,q))$ are not on the quadrant of $l(p,q)(x_p \neq x_q, y_p \neq y_q)$, or all points of $S \setminus (S \cap l(p,q))$ are on one open half plane which is determined by the line $l(p,q)(x_p = x_q, \text{ or } y_p = y_q)$.

Two \mathcal{O} -supports of a set S is said to be opposite if their half lines meet in two distinct points.

We denote by $\mathcal{F}(S)$ the set of all connected orthogonal convex hulls of S. For $E \in \mathcal{F}(S)$, if there exist two opposite O-supports H and L of S intersecting in two distinct points, say p and q, with $x_p \neq x_q, y_p \neq y_q$, then there exists a monotone path connecting p and q in E. We define all points on such path (not including p and q) to be semi-isolated points of E. If there exists an element of $\mathcal{F}(S)$ that has no semi-isolated point then $\bigcap_{E \in \mathcal{F}(S)} E$ is a connected orthogonal convex hull of S. Therefore, $\mathcal{F}(S)$ has only one element, denoted it by COCH(S) [3]. From now on, we suppose that $\mathcal{F}(S)$ has only one element, i.e., its element COCH(S) has no semi-isolated point.

Definition 4 ([3]) Let P be a finite planar point set. We define a point $u \in \text{COCH}(P)$ to be *extreme* of COCH(P) if there exists an orthogonal line l (u is the vertex of l) whose intersection with COCH(P) is only u and there is no point of $\text{COCH}(P) \setminus \{u\}$ which lies in the quadrant determined by l. We denote all extreme points of COCH(P) briefly by o-ext(COCH(P)).

Remark 1 If |P| > 2, there exist two distinct extreme points of COCH(P). Indeed, we consider the following cases:

- If P has more than two distinct points and the points of P belong to a straight line, then the two ending points are the two distinct extreme points of COCH(P).
- If P has more than two distinct points and the points of P are not collinear, then by Proposition 3 ii), two distinct extreme points of COCH(P) are chosen as the highest leftmost and the lowest rightmost, or the leftmost highest and the rightmost lowest, or the rightmost highest and the leftmost lowest, or the highest rightmost and the lowest leftmost (see the definition of these points in Definition 9).

Lemma 1 ([3]) We have o-ext(COCH(P)) $\subset P$.

Let $u=(x_u,y_u), v=(x_v,y_v) \in S \subset \mathbb{R}^2$, L_1 norm is determined by $||u-v||_1=|x_u-x_v|+|y_u-y_v|$. We use the definition L_1 norm in Proposition 1 and Lemma 2.

Proposition 1 ([3]) Let $S \subset \mathbb{R}^2$. Then, S is connected orthogonal convex iff for all $a, b \in S$, there exists a shortest path $SP(a, b) \subset S$ joining a and b with L_1 norm, and the length of SP(a, b) is $||a - b||_1$. In addition, SP(a, b) is an increasingly monotone path (i.e., for $u, v \in SP(a, b)$, $(x_u - x_v)(y_u - y_v) \geq 0$).

The minimum rectilinear rectangle of a planar point set is a minimum rectangle having edges parallel to x or y axises that contains the set.

Lemma 2 Every connected orthogonal convex hull of a finite planar point set is included in the minimum rectilinear rectangle of the point set.

Proof. Let E be a connected orthogonal convex hull of finite planar point set P, R be the minimum rectilinear rectangle bounded of P. Assume, on contrary, $E \nsubseteq R$. Let $F = R \cap E \subsetneq E$. For all $a, b \in F$, $a, b \in E$. By orthogonal convexity of E, Proposition 1 yields that there exists a E1 shortest path E2 joining E3 and E4 such that E5 Since E6. Since E8, E7 is a monotone path in E9 and therefore, E9 is connected. We are in position to prove that E9 is orthogonal convex. Let E9 be an arbitrary horizontal line intersecting E9 (the case of E9 being a vertical line is similar). Then E9 be an arbitrary horizontal line intersecting E9 (the case of E9 being a vertical line is similar). Then E9 convex. Since E9 horizontal line intersection of E9 convex. Therefore, E9 is connected orthogonal convex and E9. This contradicts the fact that E9 is a smallest connected orthogonal convex hull of E9. Hence, E9 convex hull of E9. Hence, E9 convex hull of E9.

Note that Proposition 1 and Lemma 2 imply the following Proposition 2 and Proposition 2 is used in Subsection 5.2.

Proposition 2 Let $P := \{p_1, \ldots, p_m\} \subset \mathbb{R}^2$. Then, every connected orthogonal convex hull of P is compact.

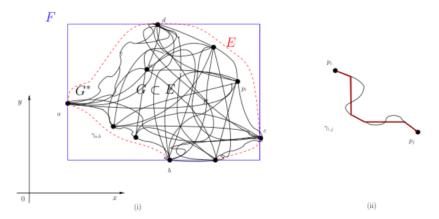


Fig. 1 (i) $G := \bigcup_{(i,j) \in [1,m] \times [1,m]} \gamma_{i,j}$ and the region G^* formed by the path γ . (ii) Adapting $\gamma_{i,j}$ to get an orthogonal convex set.

Proof. Let E be a connected orthogonal convex hull of P and F be the minimum rectangle having edges parallel to coordinates axises, where F is formed by $a,b,c,d\in P$. (see Fig. 1 (i)).

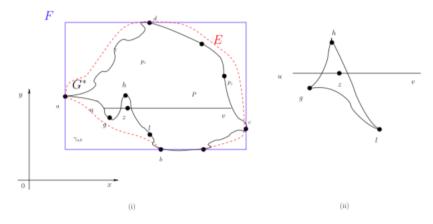


Fig. 2 $z \in [u, v]$ in the region G^* formed by the monotone paths $\gamma_{h,g}, \gamma_{h,l}$ and $\gamma_{g,l}$.

We now claim that there is a compact orthogonal convex subset set of E containing P. By Proposition 1, for each pair $(i,j) \in [1,m] \times [1,m]$, exists a staircase path belonging to E, say $\gamma_{i,j}$, joining p_i and p_j (see Fig. 1 (ii)).

Lemma 2 implies that $E \subset F$, then

$$G:=\bigcup_{(i,j)\in[1,m]\times[1,m]}\gamma_{i,j}\subset F.$$

By the way, the closedness of $\gamma_{i,j}$ yields that G is closed. Thus G is compact. Let γ be the boundary of G. We prove that γ is a path. Let $\beta_1(x) := \min\{\gamma_{ij}(x) : (i,j) \in [1,m] \times [1,n]\}$, where $x \in [a_x,b_x]$. Since minimum of finite continuous functions is also continuous, $\beta_1(x)$ is continuous. Therefore, the part of γ from a to b is a path. By similar argument,

$$\beta_2(x) := \min\{\gamma_{ij}(x) : (i,j) \in [1,m] \times [1,n]\}, \text{ where } x \in [b_x, c_x];$$

$$\beta_3(x) := \max\{\gamma_{ij}(x) : (i,j) \in [1,m] \times [1,n]\}, \text{ where } x \in [b_x, c_x];$$

$$\beta_4(x) := \max\{\gamma_i(x) : (i,j) \in [1,m] \times [1,n]\}, \text{ where } x \in [a_x, b_x]$$

are also continuous. Therefore, the parts of γ from b to c (β_2), from c to d (β_3) and from d to a (β_4) are paths. Then γ is a path. By the way chosen each part of γ , we get γ is not self-cross. Thus γ bounds a region G^* . We have $G^* \subset E$ and G^* is connected and contains P.

We are in position to prove that G^* is orthogonal convex. Take a rectilinear line k intersecting G^* . Let $u,v\in k\cap G^*$ being two "farthest" points which still lie in G^* . Assume without loss of generality that [u,v] is parallel to x-axis. We claim that $[u,v]\subset G^*$. Assume the contrary that $z\in [u,v]\setminus G^*$ (see Fig. 2 (i)). Consider the case u belongs to the part $\gamma_{a,b}$ of γ between a and b and $a_x < z_x < b_x$ (the other cases are similar). As $\gamma_{a,b}$ is formed by some monotone paths joining two points of P, there are three points $g,h,l\in P$ such that h is above $[u,v],\ g,l$ are under $[u,v],\ \gamma_{h,g}$ and $\gamma_{h,l}$ are monotone (see Fig. 2 (ii)). Since g,l are under [u,v], the monotone path $\gamma_{g,l}$ is under [u,v]. Therefore z belongs to the region formed by $\gamma_{g,l},\ \gamma_{h,g}$ and $\gamma_{h,l}$. This implies that $z\in G^*$, a contradiction. Thus, G^* is orthogonal convex.

Because E is the smallest connected orthogonal convex set containing P, we conclude that $G^* = E$. Thus E is compact.

The compactness of the connected orthogonal convex hull of a finite planar point set P is also shown in the Lemma 3 below.

A rectilinear polygon is a simple polygon whose edges are rectilinear (i.e., they are parallel to either x or y axis). The polygon has therefore only 90 and 270 degree internal angles. An (x,y)-polygon is one of the following: a) a point; b) connected rectilinear line segments; c) a rectilinear polygon; and d) a connected union of type b) and or type c) (x,y)-polygons (see [11]).

Lemma 3 ([3]) The connected orthogonal convex hull of a finite planar point set P is an orthogonal convex (x,y)-polygon whose boundary is union of finite set of O-supports, and each O-support goes through two extreme points of COCH(P).

3 Directed orthogonal lines and some related concepts

In this content we present definition a *directed orthogonal line* and some other properties necessary to serve the following section.

Given an ordered triple of points (a, b, c) in \mathbb{R}^2 , let

orient
$$(a, b, c) = \begin{vmatrix} 1 & x_a & y_a \\ 1 & x_b & y_b \\ 1 & x_c & y_c \end{vmatrix}$$
. (1)

Definition 5 ([6], p.10) We say that

- (i) The ordered triple (a, b, c) has positive orientation (negative orientation, zero orientation, resp.) if $\operatorname{orient}(a, b, c) > 0$ (orient(a, b, c) < 0, orient(a, b, c) = 0, resp.).
- (ii) The point c is called on the left of (on the right of, on, resp.) the directed line ab if orient(a, b, c) > 0 (orient(a, b, c) < 0, orient(a, b, c) = 0, resp.).

Definition 6 Let $l^v(a, b)$ be the orthogonal line through two points a and b with its vertex v. If b is on the right of av then we call $l^v(a, b)$ the directed orthogonal line from a to b and denoted it by $\mathcal{L}^v(a, b)$ (Figure 3).

Definition 7 Let $\mathcal{L}^{v}(a, b)$ be a directed orthogonal line from a to b with its vertex v. A point p is called *is on the right of* $\mathcal{L}^{v}(a, b)$ if p is on the right of both av and vb. A point p is called *is on the left of* $\mathcal{L}^{v}(a, b)$ if p is either on the left of av or on the left of vb.

We denote by P_{ab} the set containing all points of P being on the right of $\mathcal{L}^{v}(a,b)$.

Definition 8 Let $\mathcal{L}^v(a,b)$ be a directed orthogonal line from a to b with its vertex v and $p \in P_{ab}$. We call the length of [p,v] the *orthogonal distance* from p to $\mathcal{L}^v(a,b)$, denoted by $\mathrm{Odist}(p,\mathcal{L}^v(a,b))$. The point p is called the *farthest point* of P to $\mathcal{L}^v(a,b)$ if p satisfies

$$\mathrm{Odist}(p,\mathcal{L}^v(a,b)) = \max_{q \in P_{ab}} \{ \mathrm{Odist}(q,\mathcal{L}^v(a,b)) \}.$$

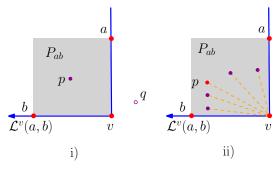


Fig. 3 i) p is on the right of $\mathcal{L}^{v}(a, b)$, q is on the left of $\mathcal{L}^{v}(a, b)$; ii) p is the farthest point to $\mathcal{L}^{v}(a, b)$.

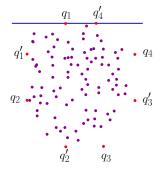


Fig. 4 Eight extreme points q_1 , q'_1 , q_2 , q'_2 , q_3 , q'_3 , q_4 , q'_4 .

Definition 9 ([8]) Let P be a finite planar point set. The point with the maximal y-coordinate (minimal y-coordinate, respectively) among the points of P having the minimal x-coordinate (maximal x-coordinate, respectively) is called the highest leftmost point (the lowest leftmost point, respectively). Similarly, we define seven other special points of P: leftmost highest, leftmost lowest, rightmost lowest, lowest rightmost, highest rightmost, rightmost highest points.

In Fig. 4, q_1 is the leftmost highest, q'_1 is the highest leftmost, q_2 is the lowest leftmost, q'_2 is the leftmost lowest, q_3 is the rightmost lowest, q'_3 is the lowest rightmost, q_4 is the highest rightmost, and q'_4 is the rightmost highest.

It is clear that the eight points in Definition 9 are the extreme points of COCH(P).

Given a point $p(x_p, y_p)$. The four orthants $o_1(p), o_2(p), o_3(p)$ and $o_4(p)$ are determined by the closed regions

$$\begin{split} o_1(p) &:= [x_p, +\infty) \times [y_p, +\infty), \\ o_2(p) &:= (-\infty, x_p] \times [y_p, +\infty), \\ o_3(p) &:= (-\infty, x_p] \times (-\infty, y_p], \\ o_4(p) &:= [x_p, +\infty) \times (-\infty, y_p] \end{split}$$

as the orthants of the point p (see Fig. 5).

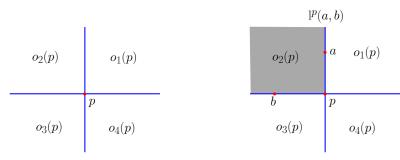


Fig. 5 $o_1(p), o_2(p), o_3(p)$ and $o_4(p)$ are the orthants of the point p.

Fig. 6 $o_2(p)$ (shaded area) coincides with the quadrant defined by orthogonal line $l^p(a, b)$.

Definition 10 (see [5]) A point $p \in P$ is called *maximal point* if at least one its orthant does not contain any points of $P \setminus \{p\}$.

Remark 2 It is easy to see that four orthants of a point coincide with the quadrants defined by four orthogonal lines of the same point.

Lemma 4 If $p \in COCH(P)$ then each orthant of p contains at least an extreme point of COCH(P).

Proof. Taking a point $p \in COCH(P)$, we consider three cases: p is an extreme point of COCH(P), p is on the boundary but is not an extreme point of COCH(P) (see Fig. 7) and p is in the interior of COCH(P) (see Fig. 8).

Case 1: p is an extreme point of COCH(P). Then clearly all four orthants of p contain the extreme point p of COCH(P).

Case 2: p is on the boundary but is not an extreme point of COCH(P). According to Lemma 3, p must belong to an \mathcal{O} -support l(a,b) that goes through two extreme points a and b of COCH(P). If \mathcal{O} -support l(a,b) is a straight line (i.e, $x_a = x_b$ or $y_a = y_b$), then clearly two orthants of p contain the extreme point a and the remaining two orthants contain the extreme point b (see Figure 7 i)). As two some orthants of p contain the extreme point a, without loss of generality assume that p belongs to the rectilinear half line containing b(see 7 ii)). Then two orthants of p contain the extreme point b. One of the remaining two orthants contains the extreme point a. We will show that the final orthants of p (e.g., $o_4(p)$ as shown in Fig. 7 ii)) contain at least one other extreme point of COCH(P). Indeed, suppose that $o_4(p)$ does not contain any extreme point of COCH(P). Then, by Lemma 3, $o_4(p)$ does not contain any point of P. Let c be a smallest y-coordinate point among the points of P in $o_2(p)$, and d be a greatest x-coordinate point among the points of P in $o_3(p)$ (because P is finite, such c and d exist). Then an orthogonal line l(c,d) is an O-support and the intersection of l(c, d) and l(a, b) consists of two distinct points. Therefore, connected orthogonal convex hulls of P have semi-isolated points. This is contrary to assumption (A).

Case 3: p is in the interior of COCH(P). Then the orthants $o_1(p)$, $o_2(p)$, $o_3(p)$, and $o_4(p)$ intersect $COCH(P) \setminus \{p\}$. Applying the similar argument for $o_4(p)$ in Case 2 to these orthants, we conclude that each orthant contains at least one extreme point of COCH(P).

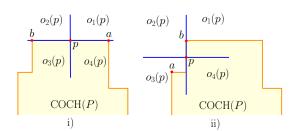


Fig. 7 p lies on the boundary of COCH(P).

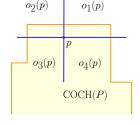


Fig. 8 p is inside the boundary point of COCH(P).

We recall the following result introduced in [5].

Lemma 5 (see [5]) Let P be the set of n points chosen according to any probability distribution Δ . Then the expected number of maximal is $O(\log n)$.

We denote the set of maximal points of P by $\mathcal{M}(P)$.

Lemma 6 Let P be a finite planar point set satisfying (A). Then

$$\mathcal{M}(P) = \text{o-ext}(\text{COCH}(P)).$$

Proof. Let $p \in \mathcal{M}(P)$, according to Definition 10, at least one of its orthants does not contain any points of $P \setminus \{p\}$. Without loss of generality we assume that $o_2(p)$ does not contain any points of $P \setminus \{p\}$. We will prove $p \in \text{o-ext}(\text{COCH}(P))$, that is, according to Definition 4 and Remark 2, we will prove that the intersection of $o_2(p)$ and COCH(P) is only u and there is no point of $\text{COCH}(P) \setminus \{p\}$ which lies in $o_2(p)$. We will prove this by contradiction. Indeed, suppose that there exists $u \in o_2(p) \cap \text{COCH}(P)$ and $u \neq p$. There are two following cases:

- u is an extreme point of COCH(P). According to Lemma 1, $u \in P$, namely $o_2(p)$ contains a points $u \in P \setminus \{p\}$.
- -u is not an extreme point of $\mathrm{COCH}(P)$. According to Lemma 4, each orthant of u contains at least one extreme point of $\mathrm{COCH}(P)$. Therefore, $o_2(u)$ also contains at least one extreme point, say t, of $\mathrm{COCH}(P)$ and according to Lemma 1, $t \in P$. Furthermore, since $u \in o_2(p)$, we have $o_2(u) \subset o_2(p)$. It follows that $t \in o_2(p)$.

Both cases above contradict the hypothesis that $o_2(p)$ does not contain any points of $P \setminus \{p\}$. Hence, $p \in \text{o-ext}(\text{COCH}(P))$. Therefore

$$\mathcal{M}(P) \subset \text{o-ext}(\text{COCH}(P)).$$
 (2)

Conversely, let $p \in \text{o-ext}(\text{COCH}(P))$, we are in position to prove that $p \in \mathcal{M}(P)$. Indeed, Lemma 1 implies $p \in P$. According to Definition 4 and remark 2, there exists an orthant o(p) of p that does not contain any point of COCH(P). Therefore o(p) also does not contain any points of P. It follows that the point $p \in \mathcal{M}(P)$. Therefore

$$o\text{-ext}(COCH(P)) \subset \mathcal{M}(P).$$
 (3)

It follows from (2) and (3) that
$$\mathcal{M}(P) = \text{o-ext}(\text{COCH}(P))$$
.

Let P be a finite planar point set. The following proposition is needed to prove the correctness of the algorithm in the next section.

Proposition 3 Let p, q $(x_p \neq x_q, y_p \neq y_q)$ be any two distinct extreme points of COCH(P) and r be a farthest point of P_{pq} from the directed orthogonal line $\mathcal{L}^v(p,q)$. Then

- i) r is an extreme point of COCH(P).
- ii) For each $w \in o(r)_v \cap P_{pq}$, $\operatorname{dist}(w, \operatorname{vq}) \leq \operatorname{dist}(p, v)$ and $\operatorname{dist}(w, \operatorname{vp}) \leq \operatorname{dist}(q, v)$, we have $o(r)_v \cap o(w)_v \cap P_{pq} \cap \operatorname{o-ext}(\operatorname{COCH}(P)) = \emptyset$, where $o(w)_v$ is an orthant of w and contains v. Consequently, w is not an extreme point of $\operatorname{COCH}(P)$.

Proof. Consider the case $x_p > x_q$ and $y_p > y_q$ (the other cases are similar). i) We claim that

$$o_2(r)$$
 does not contain any point of $P_{pq} \setminus \{r\}$. (4)

Assume the contrary that there is a point $t \in o_2(r) \cap (P_{pq} \setminus \{r\})$. Because $\angle(tr, vr) \ge \pi$, we get $\operatorname{dist}(v, t) > \operatorname{dist}(v, r)$ and therefore r is not a farthest point of P_{pq} , a contradiction. Thus (4) holds true and therefore r is an extreme point of $\operatorname{COCH}(P)$.

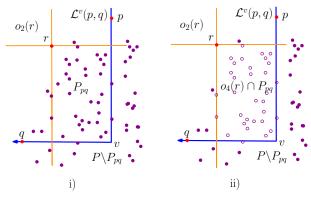


Fig. 9 i) The farthest point r to $\mathcal{L}^v(p,q)$ is an extreme point of $\mathrm{COCH}(P)$; ii) All the points in the set $o_4(r) \cap P_{pq} \setminus \{r\}$ are not the extreme points of $\mathrm{COCH}(P)$.

ii) Take $w1 \in o(r)_v \cap o(w)_v \cap P_{pq}$. Then, $r \in o_2(w1)$, $p \in o_1(w1)$ (as $\operatorname{dist}(w1, vq) \leq \operatorname{dist}(w, vq) \leq \operatorname{dist}(p, v)$), $q \in o_3(w1)$ (as $\operatorname{dist}(w1, vp) \leq \operatorname{dist}(w, vp) \leq \operatorname{dist}(q, v)$), and $v \in o_4(w)$. By the definition of extreme points of $\operatorname{COCH}(P)$, w1 is not an extreme point of $\operatorname{COCH}(P)$. Choose w1 = w, we conclude that w is not an extreme point of $\operatorname{COCH}(P)$.

Remark 3 Among the points of the set P_{pq} , there can be more than one farthest point from the directed orthogonal line $\mathcal{L}^v(p,q)$. However, all of them are the extremes of COCH(P) (according to Proposition 3).

4 Algorithm based on Quickhull for finding the connected orthogonal convex hull

4.1 \mathcal{O} -Quickhull algorithm

Consider the following four cases: the leftmost highest concides with the highest leftmost, and the lowest leftmost concides with the leftmost lowest, and the rightmost lowest concides with the lowest rightmost, and the highest rightmost concides with the rightmost highest, COCH(P) is a rectangle formed by

these points. That is reason why we can assume from now on that at least one of the cases above does not hold.

Inspired by the idea of the Quickhull algorithm [4,6,7,13], we now present a new efficient algorithm, namely $\mathcal{O}\text{-Quickhull}$, for finding the connected orthogonal convex hull COCH(P) of a finite planar point set P under the assumption (A). The first step of the $\mathcal{O}\text{-Quickhull}$ algorithm is to find two distinct extreme points, say a and b, of COCH(P) (this is always guaranteed according to Remark 1). Let $\mathcal{L}^s(a,b)$ be the directed orthogonal line with its vertex s from a to b. Note that, P_{ab} the set containing all points on the right of $\mathcal{L}^s(a,b)$. Then, from P_{ab} find the farthest point, say c, from the directed orthogonal line $\mathcal{L}^s(a,b)$. Add the point c to o-ext(COCH(P)). Let $\mathcal{L}^{s_1}(a,c)$ ($\mathcal{L}^{s_2}(c,b)$, resp.) be the directed orthogonal line with its vertex s_1 (s_1 , resp.) from a to c (from c to b). To find the next extreme points of COCH(P), we replace the directed orthogonal line $\mathcal{L}^s(a,b)$ by $\mathcal{L}^{s_1}(a,c)$ and $\mathcal{L}^{s_2}(c,b)$, and recursively continue the algorithm.

 \mathcal{O} -Quickhull algorithm illustrates the function \mathcal{O} -Quickhull (a, b, P_{ab}) , where a, b is two distinct extreme points of $\mathrm{COCH}(P)$ and P_{ab} is the set of all points on the right of the directed orthogonal line $\mathcal{L}^s(a, b)$ with its vertex s from a to b. The output of \mathcal{O} -Quickhull algorithm is the subset of P_{ab} containing all the extreme points of $\mathrm{COCH}(P)$. We use " \cup " to represent list concatenation. The final orthogonal convex hull is found when we choose pairs of extreme points to apply \mathcal{O} -Quickhull algorithm (see Subsection 5.2).

Proposition 3 ii) allows us not to consider points $w \in o(r)_v \cap P_{pq}$, $\operatorname{dist}(w, vq) \leq \operatorname{dist}(p, v)$ and $\operatorname{dist}(w, vp) \leq \operatorname{dist}(q, v)$, here p, q, r play the roles of a, b, c, respectively. To do so at every step, we choose initial distinct points a and b, respectively such that $\operatorname{dist}(c, sb) \leq \operatorname{dist}(a, s)$ and $\operatorname{dist}(c, sa) \leq \operatorname{dist}(b, s)$.

Algorithm 1 O-Quickhull algorithm

function \mathcal{O} -Quickhull (a, b, P_{ab})

- 1. If $P_{ab} = \emptyset$ then return ()
- 2. else
 - (a) $c \leftarrow$ the farthest point from $\mathcal{L}^s(a,b)$.
 - (b) $P_{ac} \leftarrow$ the set of points on the right of the directed orthogonal line $\mathcal{L}^{s_1}(a,c)$ from a to c with its vertex s_1 .
 - (c) $P_{cb} \leftarrow$ the set of points on the right of the directed orthogonal line $\mathcal{L}^{s_2}(c,b)$ from c to b with its vertex s_2 .
 - (d) return \mathcal{O} -Quickhull $(a, c, P_{ac}) \cup \{c\} \cup \mathcal{O}$ -Quickhull (c, b, P_{cb}) .

4.2 The correctness and the complexity of \mathcal{O} -Quickhull algorithm

Next we will prove the correctness of $\mathcal{O}\textsc{-Quickhull}$ algorithm in the following Theorem 1.

Theorem 1 The output of \mathcal{O} -QUICKHULL algorithm is the subset of P_{ab} containing all the extreme points of COCH(P).

Proof. Because the function \mathcal{O} -Quickhull (a, b, P_{ab}) chooses the farthest point c from $\mathcal{L}^s(a, b)$, c is an extreme point of $\mathrm{COCH}(P)$ (Proposition 3 i)). Thus, \mathcal{O} -QUICKHULL detects points in $P_{ac} \cup P_{cb}$ and the output consists of extreme points of $\mathrm{COCH}(P)$. Because initial distinct points a and b are such that $\mathrm{dist}(c, \mathrm{sb}) \leq \mathrm{dist}(a, \mathrm{s})$ and $\mathrm{dist}(c, \mathrm{sa}) \leq \mathrm{dist}(b, \mathrm{s})$, we get that $o(c)_s \cap P_{ab}$ does not contain any extreme points of $\mathrm{COCH}(P)$ (Proposition 3 ii)). It follows that the output is the set of all extreme points of $\mathrm{COCH}(P)$ in P_{ab} .

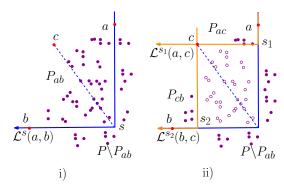


Fig. 10 i) The set P_{ab} with the farthest points c from $\mathcal{L}^s(a,b)$; ii) The set P_{ac} .

We will discuss the following simple analysis of the time complexity of $\mathcal{O}\textsc{-Quickhull}$ algorithm.

Remark 4 In \mathcal{O} -Quickhull algorithm, after finding the extreme point c, in order to find the next extreme points, we only need to consider the points in the set P_{ac} and P_{cb} , that is, the points in the set $P_{ab} \setminus (P_{ac} \cup P_{cb})$ are not considered anymore because these points cannot be the extreme points of the COCH(P) (see Fig. 10 ii) and Proposition 3 ii)). Thus, the number of points to detect the next extreme point will decrease significantly.

Theorem 2 Suppose that the set P_{ab} consists n points. The worst case complexity of the \mathcal{O} -Quickhull algorithm is $O(n^2)$ and its expected complexity is $O(n \log n)$.

Proof. Suppose that the output of \mathcal{O} -Quickhull algorithm has m extreme points of $\mathrm{COCH}(P)$. The algorithm calls \mathcal{O} -Quickhull functions (m+1) times. In which, each of the first m functions finds exactly one extreme point of $\mathrm{COCH}(P)$ and need O(n) time complexity. The last time work with an empty set. So the time complexity of \mathcal{O} -Quickhull algorithm is O(mn). In the worst case, when m=n (here n is the number of points of P_{ab}), we have

the worst time complexity of $O(n^2)$. According to Lemma 5 and Lemma 6, the expected number of extreme points of $\mathrm{COCH}(P)$ in P_{ab} is $O(\log n)$, i.e., $m = O(\log n)$. Therefore, the expected complexity of \mathcal{O} -Quickhull algorithm is $O(n \log n)$.

5 Implementation

In this section, we are going to compare the running times of our algorithm to \mathcal{O} -Graham introduced by An, Huyen and Le in [3] and an other algorithm proposed by Montuno and Fournier in [11].

5.1 The Test Sets

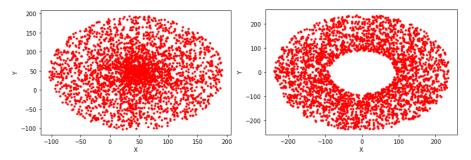


Fig. 11 Disc data with 3000 points.

Fig. 12 Hollow disc data with 3000 points.

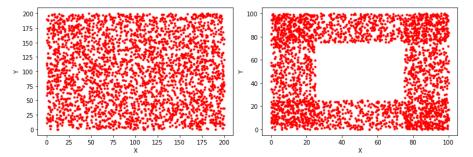
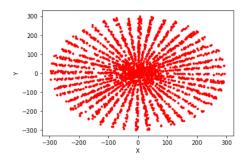


Fig. 13 Square data with 3000 points.

 ${\bf Fig.~14~~Hollow~square~data~with~3000~points}.$

To test the algorithms we create five data types, below are specific descriptions for these 6 data types



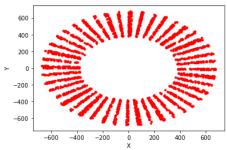


Fig. 15 Sun data with 3000 points.

Fig. 16 Hollow Sun data with 3000 points.

- Disc data: We generate random real points in a disc. For instance, the input data in Fig. 11 consists of 3000 points.
- Hollow disc data: We create two concentric discs (or ellipses) of different radii. The input points are created randomly outside the smaller disc (or ellipse) but inside the bigger one. Different examples are created with different radii ratios (see Fig. 12).
- Square data: The input points are randomly generated inside a square (see Fig. 13).
- Hollow square data: The points are created randomly inside a square and outside another smaller concentric square. We create data corresponding to different sizes of the smaller square (see Fig. 14).
- Sun data: The points are randomly generated according to the central angle of a disc and interspersed equal angles with no points (see Fig. 15).
- Hollow sun data: The points are randomly generated according to the central angle of a disc and interspersed equal angles with no points. In addition, these points are also created outside the concentric disc with the one above (see Fig. 16).

5.2 Numerical Results

In this subsection we present the selection of pairs of extreme distinct points a and b such that $\operatorname{dist}(c, \operatorname{sb}) \leq \operatorname{dist}(a, \operatorname{s})$ and $\operatorname{dist}(c, \operatorname{sa}) \leq \operatorname{dist}(b, \operatorname{s})$ hold true to apply $\mathcal{O}\text{-Quickhull}$ algorithm to find the final orthogonal convex hull of a finite set of a planar points.

It is known that the points lying inside or on the edges of the polygon which its edges are parallel to x-axis or y-axis connecting eight extreme points $q_1, q'_1, q_2, q'_2, q_3, q'_3, q_4$ and q'_4 (except the eight these points) are not extreme points of COCH(P) and can be deleted (see Figure 17).

Due to the compactness of COCH(P), we can determine its boundary according to Lemma 3 as the union of finite set of O-supports, where each O-support goes through two extreme points of COCH(P). Assume that $q_1 \neq q'_1$ and $p_0 = q_1, p_1, \ldots, p_{k-1}, p_k = q'_1$ are such extreme points of COCH(P) on the the right of the directed orthogonal line from q_1 to q'_1 . They form a staircase

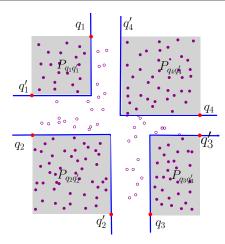


Fig. 17 Four sets $P_{q_i q'_i}$, i = 1, 2, 3, 4.

path, say, $\mathcal{P}_{q_1q_1'}$ (i.e., a union of parts of O-supports through p_i and p_{i+1} , $i=0,\ldots,k-1$) joining q_1 and q_1' . Similarly, we can also determine $\mathcal{P}_{q_iq_i'}$, i=2,3,4. Thus COCH(P) is an orthogonal convex (x,y)-polygon whose boundary is union of the rectilinear line segments $[q_i,q_i']$, i=1,2,3,4 and staircase paths $\mathcal{P}_{q_iq_i'}$, joining q_i and q_i' , i=1,2,3,4, respectively.

To find staircase paths $\mathcal{P}_{q_iq_i'}$, i=1,2,3,4, we apply \mathcal{O} -QUICKHULL algorithm for set $P_{q_1q_1'}$ if $q_1 \neq q_1'$, set $P_{q_2q_2'}$ if $q_2 \neq q_2'$, set $P_{q_3q_3'}$ if $q_3 \neq q_3'$, set $P_{q_4q_4'}$ if $q_4 \neq q_4'$. Therefore, the final orthogonal convex hull COCH(P) is

$$\{q_1\} \cup \mathcal{O}$$
-Quickhull $(q_1, q'_1, P_{q_1q'_1}) \cup \{q'_1, q_2\} \cup \mathcal{O}$ -Quickhull $(q_2, q'_2, P_{q_2q'_2})$

$$\cup \{q'_2, q_3\} \cup \mathcal{O}$$
-Quickhull $(q_3, q'_3, P_{q_3 q'_3}) \cup \{q'_3, q_4\} \cup \mathcal{O}$ -Quickhull $(q_4, q'_4, P_{q_4 q'_4})$.

If no case occurs (i.e., $q_1 = q_1'$, $q_2 = q_2'$, $q_3 = q_3'$, $q_4 = q_4'$), the rectangle $q_1q_2q_3q_4$ is the orthogonal convex hull to look for.

The algorithms are implemented in python and run on PC 1.8 GHz Intel Core i5 with 8 GB RAM. The Fig. 18- 23 illustrate the results of finding orthogonal convex hull of the sets of points corresponding to the data sets.

Tables 1 - 6 list the running times (in seconds) of the three algorithms: \mathcal{O} -Quickhull algorithm, \mathcal{O} -Graham introduced by An, Huyen and Le in [3] (\mathcal{O} -Graham, in short) and the algorithm proposed by Montuno and Fournier in [11] (Montuno and Fournier 's algorithm, in short).

In general, the \mathcal{O} -Quickhull algorithm runs faster than the others. For convenience, in Table 7, we list the ratios of the actual running times of \mathcal{O} -graham algorithm and Montuno and Fournier's algorithm to the \mathcal{O} -Quickhull algorithm. In the last row of Table 7, we compute the ratios of all the tested data for the overall results. All ratios are calculated using the geometric mean. The average result on all the data, our algorithm is 1.431 times faster than \mathcal{O} -Graham algorithm and 1.401 times faster than Montuno and Fournier's algorithm. To explain this, after finding an extreme point of COCH(P), a large

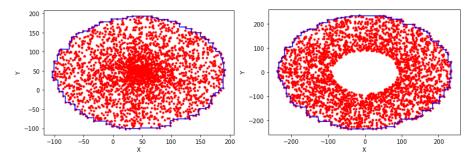


Fig. 18 Disc data with 3000 points.

Fig. 19 Hollow disc data with 3000 points.

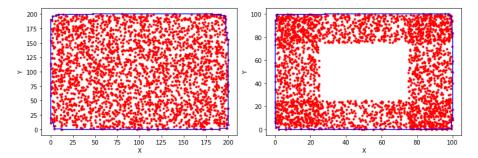


Fig. 20 Square data with 3000 points.

Fig. 21 Hollow square data with 3000 points.

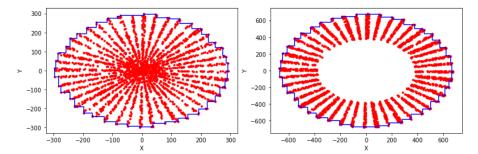
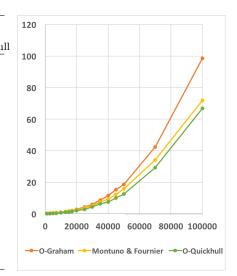


Fig. 22 Under triangle with 3000 points.

Fig. 23 Hollow Sun data with 3000 points.

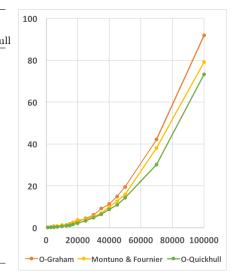
number of points that are certainly not the extreme points of $\mathrm{COCH}(P)$ is ignored, i.e, the number of points to consider when finding a new extreme point will reduce significantly (see Remark 4). Furthermore, our algorithm does not need a preprocessing step, which rearranges the input points as \mathcal{O} -graham algorithm and Montuno and Fournier's algorithm.

Montuno and				
	Input	O-	Fournier 's	O-
		Graham	algorithm	Quickhu
	1000	0.18	0.332	0.138
	3000	0.259	0.473	0.187
	5000	0.362	0.625	0.245
	7000	0.685	0.777	0.344
	10000	0.999	1.115	0.594
	13000	1.333	1.379	0.993
	15000	1.724	1.868	1.131
	17000	2.096	1.964	1.418
	20000	2.746	2.363	2.012
	25000	4.420	3.940	2.861
	30000	5.868	5.140	4.331
	35000	8.635	7.503	6.321
	40000	11.532	9.143	7.558
	45000	15.217	12.099	9.939
	50000	18.504	15.844	12.590
	70000	42.312	34.126	29.271
	100000	98.457	71.852	66.852



 ${\bf Table~1}~{\bf The~actual~running~times~(in~seconds)~of~the~algorithms~for~disc~data}.$

		Montuno and	
Input	O-	Fournier 's	O-
	Graham	algorithm	Quickhu
1000	0.144	0.280	0.127
3000	0.201	0.446	0.189
5000	0.429	0.798	0.258
7000	0.497	0.873	0.351
10000	0.914	1.224	0.635
13000	1.298	1.444	0.929
15000	1.839	1.763	1.145
17000	2.132	2.652	1.597
20000	2.923	3.687	2.139
25000	4.471	4.059	3.280
30000	6.073	5.166	4.820
35000	9.167	6.918	6.411
40000	11.373	9.940	8.744
45000	14.945	12.750	10.934
50000	19.502	15.873	14.236
70000	42.137	38.068	30.196
100000	91.748	79.085	73.189



 ${\bf Table~2~~ The~actual~ running~ times~(in~seconds)~ of~the~algorithms~for~hollow~disc~data}.$

6 Concluding Remarks

We have provided an efficient algorithm, based on the idea of quickhull, for finding the connected orthgonal convex hull of a finite planar point set and have compared it with the algorithm [3] and Montuno and Fournier's algorithm [11]. A similar algorithm for finding \mathcal{O}_{β} -convex hulls (introduced in [2])

		Montuno and		90	
Input	O-	Fournier 's	O-		
-	Graham	algorithm	Quickhull	80	
1000	0.278	0.230	0.147		
3000	0.247	0.288	0.188	70	
5000	0.406	0.375	0.253	60	
7000	0.493	0.451	0.318		
10000	0.896	0.646	0.557	50	
13000	1.195	1.036	0.748		
15000	1.448	1.379	1.180	40	
17000	1.932	1.742	1.335	20	
20000	2.283	2.079	1.734	30	
25000	3.556	3.264	2.738	20	
30000	5.452	4.841	4.063		
35000	7.867	6.991	5.868	10	
40000	10.376	9.561	7.698		
45000	12.003	10.478	10.033	0	
50000	17.462	16.018	13.466	0 20000 40000 60000 80000 100000	
70000	39.573	38.954	30.962	a Cookers a Mantage Country a Contribut	
100000	81.678	78.239	70.768	O-Graham Montuno & Fournier O-Quickl	

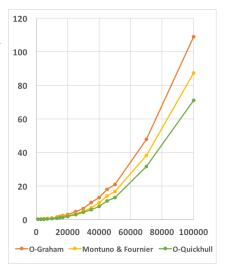
Table 3 The actual running times (in seconds) of the algorithms for square data.

		Montuno and			
Input	O-	Fournier 's	O-	90	
mpat	Graham	algorithm	Quickhull	80	
1000	0.196	0.247	0.133		
3000	0.211	0.304	0.226	70	
5000	0.331	0.371	0.237	60	
7000	0.405	0.451	0.295		
10000	0.791	0.687	0.528	50	
13000	1.117	1.034	0.803		
15000	1.411	1.212	1.027	40	
17000	1.626	1.525	1.176	20	
20000	2.234	2.112	1.769	30	
25000	3.399	3.318	2.641	20	
30000	4.987	4.991	3.978		
35000	6.363	6.544	5.630	10	
40000	8.958	9.591	7.812		
45000	10.776	11.496	9.924	0 20000 40000 60000 80000 100000	
50000	15.834	16.945	12.901	0 20000 40000 60000 80000 100000	
70000	32.323	33.717	31.130	→ O-Graham → Montuno & Fournier → O-Quickhull	
100000	75.321	76.787	70.491	5 G-Granam 5 Workland & Pourmer 5-O-Quicki	

 ${\bf Table~4~~ The~ actual~ running~ times~(in~ seconds)~ of~ the~ algorithms~ for~ hollow~ square~ data}.$

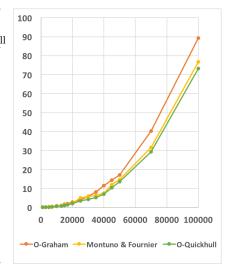
can be given. In addition, we can use the idea of space subdivision [19] to give efficient algorithms for finding such hulls. They will be the subject of another paper.

		Montuno and	
Input	O-	Fournier 's	O-
	Graham	algorithm	Quickhull
1000	0.174	0.280	0.144
3000	0.250	0.442	0.181
5000	0.348	0.554	0.239
7000	0.588	0.647	0.359
10000	0.841	0.888	0.512
13000	1.433	1.250	0.849
15000	1.867	1.555	1.093
17000	2.334	2.054	1.300
20000	2.966	2.219	1.861
25000	4.801	3.477	2.961
30000	6.661	4.958	4.360
35000	10.259	6.934	5.857
40000	13.051	10.096	7.812
45000	18.031	14.050	11.042
50000	20.918	16.801	13.147
70000	47.831	38.197	31.588
100000	108.946	87.346	70.990



 $\textbf{Table 5} \ \ \textbf{The actual running times (in seconds) of the algorithms for sun data}.$

	Montuno and		
Input	O-	Fournier 's	O-
	Graham	algorithm	Quickhu
1000	0.156	0.347	0.137
3000	0.251	0.339	0.177
5000	0.311	0.451	0.240
7000	0.442	0.681	0.327
10000	0.891	0.867	0.653
13000	1.158	1.134	0.869
15000	1.811	1.412	1.112
17000	2.026	1.725	1.407
20000	2.834	2.212	2.016
25000	4.099	5.118	3.567
30000	5.887	5.791	4.334
35000	8.163	6.440	5.241
40000	11.558	7.591	7.032
45000	14.476	11.996	10.496
50000	17.234	14.745	13.709
70000	40.323	31.717	29.320
100000	89.321	76.787	73.227



 ${\bf Table~6}~{\rm The~actual~running~times~(in~seconds)~of~the~algorithms~for~hollow~sun~data}.$

Acknowledgment

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Table 7 The ratios of the actual running times of \mathcal{O} -graham algorithm and Montuno and Fournier's algorithm to the \mathcal{O} -Quickhull algorithm.

Data types	The ratio of \mathcal{O} -Graham to \mathcal{O} -QUICKHULL	The ratio of Montuno and Fournier's algorithm to O-QUICKHULL
Disc data	1.479	1.516
Hollow disc data	1.355	1.539
Square data	1.394	1.264
Hollow square data	1.244	1.288
Sun data	1.579	1.467
Hollow sun data	1.364	1.365
All data	1.431	1.401

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