O-Quickhull: A Fast and Efficient Algorithm for Determining the Connected Orthogonal Convex Hulls

Nguyen Kieu Linh \cdot Phan Thanh An \cdot Tran Van Hoai

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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, namely $\mathcal{O}\text{-QUICKHULL}$, 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, $\mathcal{O}\text{-QUICKHULL}$ 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.

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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 [17], and VLSI circuit layout

 $E\text{-}mail: thanhan@hcmut.edu.vn, hoai@hcmut.edu.vn}$

 $^{{\}rm N.~K.~Linh^1}$

 $^{^1\}mathrm{Posts}$ and Telecommunications Institute of Technology, Hanoi, Vietnam E-mail: linhnk@ptit.edu.vn

P. T. An and T. V. $\mathrm{Hoai}^{2,3}$

 $^{^2{\}rm Ho}$ Chi Minh City University of Technology (HCMUT), 268 Ly Thuong Kiet Street, District 10, Ho Chi Minh City, Vietnam

 $^{^3{\}rm Vietnam}$ National University Ho Chi Minh City, Linh Trung Ward, Thu Duc District, Ho Chi Minh City, Vietnam

design [18], digital images processing [15]. The concept of orthogonal convex hull of a set was first mentioned in 1959 by Unger [19]. 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 the algorithm $\mathcal{O}\text{-Quickhull}$, 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, $\mathrm{dist}(p, q) := \sqrt{(x_p - x_q)^2 + (y_p - y_p)^2}$.

2.1 Connected orthogonal convex hulls

Definition 1 (see [19]) A set $S \subset \mathbb{R}^2$ is said to be *orthogonal convex* if its intersection with any horizontal or vertical line is convex.

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

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

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 (see [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 $u,v \in SP(a,b)$, $(x_u-x_v)(y_u-y_v) \geq 0$).

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 $a \neq b$ be two given points in the plane. We define $l(a,b)(x_a \neq x_b, y_a \neq y_b)$ through a, b to be union of two rectilinear half lines having the same starting point. If $x_a = x_b$ or $y_a = y_b$ then l(a,b) is the line through a and b. The set l(a,b) is called the *orthogonal line* through a and b. The common point of two the rectilinear half lines of l(a,b) is called the *vertex* of l(a,b). We also denote by $l^v(a,b)$ the orthogonal line l(a,b) having the vertex v.

An orthogonal line l(a, b), $(x_a \neq x_b, y_a \neq y_b)$ separates the plane into two regions. The quadrant region together with the orthogonal line l(a, b) will be called a *quadrant* determined by the orthogonal line and denoted by q(a, b).

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

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 \mathcal{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 $\mathrm{COCH}(S)$ [3]. From now on, we suppose that $\mathcal{F}(S)$ has only one element, i.e., its element $\mathrm{COCH}(S)$ has no semi-isolated point.

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

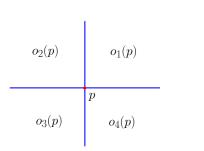
$$o_1(p) := [x_p, +\infty) \times [y_p, +\infty),$$

$$o_2(p) := (-\infty, x_p] \times [y_p, +\infty),$$

$$o_3(p) := (-\infty, x_p] \times (-\infty, y_p],$$

$$o_4(p) := [x_p, +\infty) \times (-\infty, y_p]$$

as the orthants of the point p (see Fig. 1).



 $o_2(p)$ $o_1(p)$ $o_2(p)$ $o_3(p)$ $o_4(p)$

Fig. 1 $o_1(p), o_2(p), o_3(p)$ and $o_4(p)$ are the orthants of the point p.

Fig. 2 $o_2(p)$ (shaded area) coincides with the quadrant defined by orthogonal line $l^p(a,b)$.

Remark 1 It is easy to see that four orthants of a point coincide with the quadrants defined by four orthogonal lines of the same point.

Let P be a finite planar point set. We assume that the set P satisfy that its connected orthogonal convex hulls have no semi-isolated points and call it the assumption (A).

Definition 4 (see [3]) A point $e \in COCH(P)$ is called an *extreme* (extreme point, for brevity) of COCH(P) if there exists an orthant does not contain any points of $COCH(P) \setminus \{e\}$. We denote all extreme points of COCH(P) briefly by o-ext(COCH(P)).

Incidentally, this is the same concept of maximal points of a finite planar point set P given in [5]. This can be seen later in Lemma 5.

Definition 5 An extreme point e of COCH(P) has $index\ j$ if $o_j(e),\ j = 1, 2, 3, 4$, does not contain any point of $COCH(P) \setminus \{e\}$. The set containing all the extreme points of COCH(P) with index j is denoted by $o\text{-ext}^j(COCH(P))$.

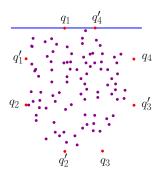


Fig. 3 Eight special points $q_1, q'_1, q_2, q'_2, q_3, q'_3, q_4, q'_4$.

2.2 Properties of connected orthogonal convex hulls of a finite planar point set

Definition 6 (see [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. 3, 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 6 are the extreme points of COCH(P).

Remark 2 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 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.

Lemma 1 (see [3]) We have

- i) o-ext(COCH(P)) $\subseteq P$.
- ii) $P \subseteq COCH(P)$.
- iii) Let P_1, P_2 be two finite point sets in the plane and $P_1 \subseteq P_2$. Then $COCH(P_1) \subseteq COCH(P_2)$.

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, $E \not\subseteq R$. 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 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 being a vertical line is similar. Then E9 convex. Since E9 here, E9 is connected orthogonal convex and E9. This contradicts the fact that E9 is a smallest connected orthogonal convex hull of E9. Hence, E1 convex hull of E9. Hence, E1 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.

The proof is given in the Appendix. 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 (see [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 \mathcal{O} -supports, and each \mathcal{O} -support goes through two extreme points of COCH(P).

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

The proof of Lemma 4 is given in the Appendix. We denote the set of maximal points of P by $\mathcal{M}(P)$.

Lemma 5 Let P be a finite planar point set satisfying (A). Then a maximal of P is an extreme point of COCH(P) and vice versa. Consequently,

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

The proof of Lemma 5 is given in the Appendix. The following lemma is needed to prove the complexity of the algorithm in Section 4.

Lemma 6 (see [5]) Let P be the set of n points chosen according to any probability distribution Δ . Then the expected number of maximal points of P is $O(\log n)$.

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 7 (see [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 8 Let $l^v(a, b)$ be the orthogonal line through two points a and b $(x_a \neq x_b, y_a \neq y_b)$ 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 4).

Definition 9 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 (Figure 4(i)).

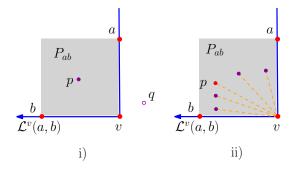


Fig. 4 i) p is on the right of $\mathcal{L}^v(a,b)$, q is on the left of $\mathcal{L}^v(a,b)$; ii) p is the farthest point to $\mathcal{L}^v(a,b)$.

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

Definition 10 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 $\text{Odist}(p, \mathcal{L}^v(a, b))$. The point p is called the *farthest point* of P_{ab} to $\mathcal{L}^v(a, b)$ if p satisfies

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

Note from Definition 5 that an extreme point e of COCH(P) has $index\ j$, j=1,2,3,4, if the orthant $o_j(e)$ does not contain any points of $COCH(P)\setminus\{e\}$ and the set containing all the extreme point of COCH(P) with index j is denoted by $o\text{-ext}^j(COCH(P))$.

Lemma 7 Let a, b $(x_a \neq x_b, y_a \neq y_b)$ be any two distinct extreme points of COCH(P). Then

$$\operatorname{o-ext}_{ab}^{j}(\operatorname{COCH}(P)) \subseteq \operatorname{o-ext}(\operatorname{COCH}(P_{ab})),$$

where o-ext $_{ab}^{j}(\text{COCH}(P))$ the set of all the extreme points with index j of COCH(P) in P_{ab} .

Proof. Let $e \in \text{o-ext}_{ab}^j(\text{COCH}(P))$. According to Definition 5, the orthant $o_j(e)$ does not contain any points of $\text{COCH}(P) \setminus \{e\}$. Because $P_{ab} \subseteq P$ and Lemma 1(iii), we get that $\text{COCH}(P_{ab}) \subseteq \text{COCH}(P)$ and therefore $o_j(e)$ does not contain any points of $\text{COCH}(P_{ab}) \setminus \{e\}$. It follows that $e \in \text{o-ext}(\text{COCH}(P_{ab}))$, i.e, $\text{o-ext}_{ab}^j(\text{COCH}(P)) \subseteq \text{o-ext}(\text{COCH}(P_{ab}))$. □

Let a, b ($x_a \neq x_b, y_a \neq y_b$) be any two distinct extreme points of COCH(P). There are four cases of two points a and b as follows (see Fig. 5).

- Case 1: $x_a > x_b$ and $y_a < y_b$;
- Case 2: $x_a > x_b$ and $y_a > y_b$;
- Case 3: $x_a < x_b$ and $y_a > y_b$;

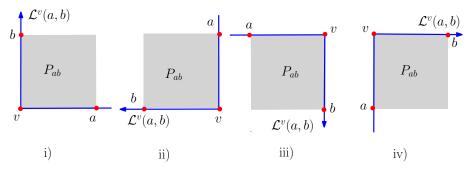


Fig. 5 The four cases of a, b: i) Case 1; ii) Case 2; iii) Case 3; iv) Case 4.

- Case 4: $x_a < x_b$ and $y_a < y_b$.

The following proposition is needed to prove the correctness of the algorithm in the next section.

Proposition 3 Let a, b ($x_a \neq x_b, y_a \neq y_b$) be any two distinct extreme points of COCH(P) and c be a farthest point of P_{ab} from the directed orthogonal line $\mathcal{L}^v(a, b)$. Then

- i) c is an extreme point with index j of COCH(P) in Case j of two points a, b, j = 1, 2, 3, 4.
- ii) $o_v(c) \cap (P_{ab} \setminus \{c\}) \cap \text{o-ext}^j(\text{COCH}(P)) = \emptyset$, where $o_v(c)$ is an orthant of c and contains v. Consequently, any point in $o_v(c) \cap (P_{ab} \setminus \{c\})$ is not an extreme point with index j of COCH(P) in P_{ab} .

Proof. Consider the Case 2 (j = 2) of two points a, b (the other cases, j = 1, 3, 4, are similar).

i) We claim that

$$o_2(c)$$
 does not contain any point of $P_{ab} \setminus \{c\}$. (2)

Assume the contrary that there is a point $t \in o_2(c) \cap (P_{ab} \setminus \{c\})$. Then we have $x_t \leq x_c$, $y_t \geq y_c$, and $t \neq c$. We get $\operatorname{dist}(t,s) > \operatorname{dist}(c,s)$ and therefore c is not a farthest point of P_{ab} from the directed orthogonal line $\mathcal{L}^s(a,b)$, a contradiction. Thus (2) holds true.

On the other hand, if a point $t \in P \setminus P_{ab}$ then $x_t \geq x_a$ or $y_t \leq y_b$. It follows that $x_t > x_c$ or $y_t < y_c$, i.e., $t \notin o_2(c)$. Therefore

$$o_2(c)$$
 does not contain any point of $P \setminus P_{ab}$. (3)

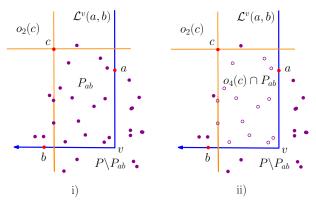


Fig. 6 In the Case 2 of a,b (i) the farthest point c to $\mathcal{L}^v(a,b)$ is an extreme point with index 2 of $\mathrm{COCH}(P)$ and (ii) all the points in the set $o_4(c) \cap (P_{ab} \setminus \{c\})$ are not the extreme points with index 2 of $\mathrm{COCH}(P)$ in P_{ab} .

From (2) and (3) we deduce that $o_2(c)$ does not contain any point of $P \setminus \{c\}$, i.e, c is an extreme point with index 2 of P. It follows from the Lemma 5 that c is an extreme point with index 2 of COCH(P).

ii) In the Case 2 of two point a and b, we get that $o_v(c) \equiv o_4(c)$. Take $d \in o_4(c) \cap (P_{ab} \setminus \{c\})$. We get that $x_c \leq x_d$ and $y_c \geq y_d$ and therefore $c \in o_2(d)$. It follows from Definition 5 that d is not an extreme point with index 2 of COCH(P) in P_{ab} .

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

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

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 $\mathcal{O}\text{-Quickhull}$ is to find two distinct extreme points, say a and b, of COCH(P) (this is always guaranteed according to Remark 2). Let $\mathcal{L}^v(a,b)$ be the directed orthogonal line with its vertex v from a to b. Note that, P_{ab} the set containing all points on the right of $\mathcal{L}^v(a,b)$. Then, from P_{ab} find the farthest point, say c, from the directed orthogonal line $\mathcal{L}^v(a,b)$. Add the point c to o-ext(COCH(P)). Let $\mathcal{L}^{v_1}(a,c)$ ($\mathcal{L}^{v_2}(c,b)$, resp.) be the directed orthogonal line with its vertex v_1 (v_2 , resp.) from a to c (from c to b). Proposition 3(ii) allows us not to consider points $t \in o_v(c) \cap (P_{ab} \setminus \{c\})$. Therefore, to find the next extreme points of COCH(P), we replace the directed orthogonal line $\mathcal{L}^v(a,b)$ by $\mathcal{L}^{v_1}(a,c)$ and $\mathcal{L}^{v_2}(c,b)$, and recursively continue the algorithm.

 \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 $\mathcal{L}^v(a,b)$ with its vertex v from a to b. If two points a, b are in Case j (j=1,2,3,4) then the output of \mathcal{O} -Quickhull contains all the extreme points with index j of $\mathrm{COCH}(P)$ in P_{ab} . 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 (see Subsection 5.2).

Algorithm 1 \mathcal{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 $\mathcal{L}^v(a, b)$.
 - (b) $P_{ac} \leftarrow$ the set of points on the right of the directed orthogonal line $\mathcal{L}^{v_1}(a,c)$ from a to c with its vertex v_1 .
 - (c) $P_{cb} \leftarrow$ the set of points on the right of the directed orthogonal line $\mathcal{L}^{v_2}(c,b)$ from c to b with its vertex v_2 .
 - (d) **return** \mathcal{O} -Quickhull $(a, c, P_{ac}) \cup \{c\} \cup \mathcal{O}$ -Quickhull (c, b, P_{cb}) .

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

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

We going to present the correctness of $\mathcal{O}\textsc{-Quickhull}$ in the following Theorem 1

Theorem 1 If two points a, b are in Case j (j = 1, 2, 3, 4) then the output of \mathcal{O} -Quickhull contains all the extreme points with index j of COCH(P) in P_{ab} .

The proof of Theorem 1 is given in the Appendix. We will discuss the following simple analysis of the time complexity of \mathcal{O} -Quickhull.

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

Proof. Suppose that the output of \mathcal{O} -Quickhull has m extreme points with index j of $\mathrm{COCH}(P)$ in P_{ab} . The algorithm calls \mathcal{O} -Quickhull functions (m+1) times. In which, each of the first m functions finds exactly one extreme point with index j 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 is O(mn). In the worst case, when m=n, we have the worst time complexity of $O(n^2)$. According to Lemma 5, Lemma 6, and Lemma 7, the expected number of extreme points with index j 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 is $O(n\log n)$.

5 Implementation

In this section, we present the selection of pairs of distinct extreme points a and b to apply \mathcal{O} -Quickhull to find the final orthogonal convex hull of a finite set of a planar points. Besides, we are going to compare the running times of \mathcal{O} -Quickhull 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

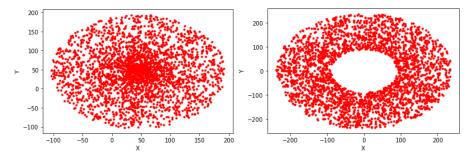


Fig. 7 Disc data with 3000 points.

Fig. 8 Hollow disc data with 3000 points.

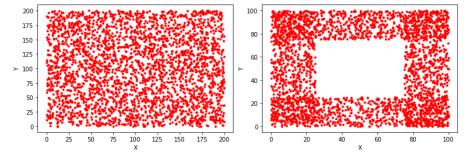
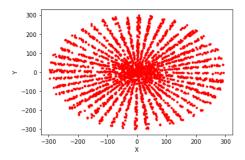


Fig. 9 Square data with 3000 points.

Fig. 10 Hollow square data with 3000 points.

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

- Disc data: We generate random real points in a disc. For instance, the input data in Fig. 7 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



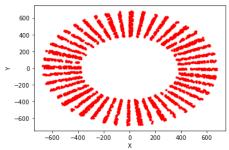


Fig. 11 Sun data with 3000 points.

Fig. 12 Hollow Sun data with 3000 points.

ellipse) but inside the bigger one. Different examples are created with different radii ratios (see Fig. 8).

- Square data: The input points are randomly generated inside a square (see Fig. 9).
- 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. 10).
- Sun data: The points are randomly generated according to the central angle of a disc and interspersed equal angles with no points (see Fig. 11).
- 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. 12).

5.2 Numerical Results

In this subsection we present the selection of pairs of distinct extreme points a and b to apply \mathcal{O} -Quickhull 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 13).

Due to the compactness of $\mathrm{COCH}(P)$, we can determine its boundary according to Lemma 3 as the union of finite set of \mathcal{O} -supports, where each \mathcal{O} -support goes through two extreme points of $\mathrm{COCH}(P)$. Thus $\mathrm{COCH}(P)$ 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 $\mathcal{P}_{q_iq_i'}$ (formed by the extreme points with the same index) joining q_i and $q_i',i=1,2,3,4$, respectively. We apply \mathcal{O} -Quickhull 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

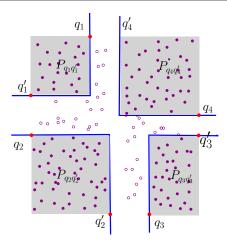


Fig. 13 Four sets $P_{q_i q'_i}$, i = 1, 2, 3, 4.

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. 14- 19 illustrate the results of finding orthogonal convex hull of the sets of points corresponding to the data sets.

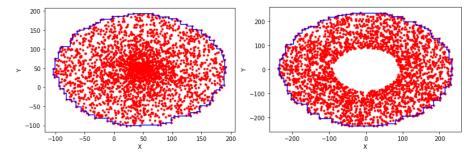


Fig. 14 The connected orthogonal convex hull of the disc data.

Fig. 15 The connected orthogonal convex hull of the hollow disc data.

Tables 1 - 6 list the running times (in seconds) of the three algorithms: \mathcal{O} -QUICKHULL, \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).

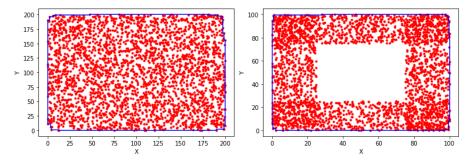


Fig. 16 The connected orthogonal convex hull of the square data.

Fig. 17 The connected orthogonal convex hull of the hollow square data.

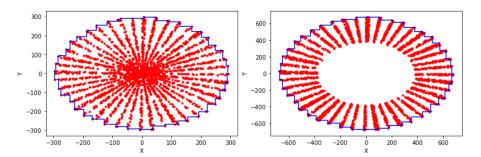
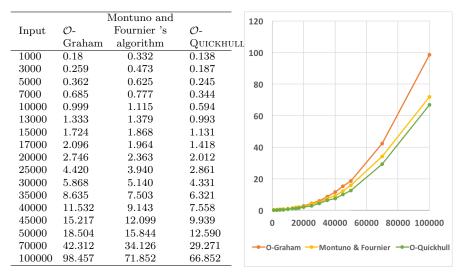


Fig. $18\,$ The connected orthogonal convex hull of the sun data.

Fig. 19 The connected orthogonal convex hull of the hollow sun data.



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

		Montuno and		100
Input	O-	Fournier 's	O-	
	Graham	algorithm	Quickhull	
1000	0.144	0.280	0.127	80
3000	0.201	0.446	0.189	80
5000	0.429	0.798	0.258	
7000	0.497	0.873	0.351	
10000	0.914	1.224	0.635	60
13000	1.298	1.444	0.929	
15000	1.839	1.763	1.145	
17000	2.132	2.652	1.597	40
20000	2.923	3.687	2.139	
25000	4.471	4.059	3.280	
30000	6.073	5.166	4.820	20
35000	9.167	6.918	6.411	20
40000	11.373	9.940	8.744	
45000	14.945	12.750	10.934	
50000	19.502	15.873	14.236	0
70000	42.137	38.068	30.196	0 20000 40000 60000 80000 100000
100000	91.748	79.085	73.189	O-Graham → Montuno & Fournier → O-Quickhull

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

		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	30
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 20000 40000 50000 00000 400000
50000	17.462	16.018	13.466	0 20000 40000 60000 80000 100000
70000	39.573	38.954	30.962	O Cychon - Blantung & Faurrian - O Ovielhu
100000	81.678	78.239	70.768	O-Graham Montuno & Fournier O-Quickhu

 ${\bf Table~3}~{\bf The~actual~running~times~(in~seconds)~of~the~algorithms~for~square~data.}$

In general, \mathcal{O} -Quickhull 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. 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. As shown in the average

		Montuno and		90
Input	O-	Fournier 's	O-	
	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	30
20000	2.234	2.112	1.769	
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	0 *************************************
45000	10.776	11.496	9.924	0 20000 40000 60000 80000 100000
50000	15.834	16.945	12.901	
70000	32.323	33.717	31.130	→O-Graham → Montuno & Fournier →O-Quickhull
100000	75.321	76.787	70.491	

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

Montuno and				120
$_{ m Input}$	O-	Fournier 's	O-	
	Graham	algorithm	Quickhuli	<u> </u>
1000	0.174	0.280	0.144	100
3000	0.250	0.442	0.181	
5000	0.348	0.554	0.239	80
7000	0.588	0.647	0.359	80
10000	0.841	0.888	0.512	
13000	1.433	1.250	0.849	60
15000	1.867	1.555	1.093	
17000	2.334	2.054	1.300	
20000	2.966	2.219	1.861	40
25000	4.801	3.477	2.961	
30000	6.661	4.958	4.360	20
35000	10.259	6.934	5.857	20
40000	13.051	10.096	7.812	
45000	18.031	14.050	11.042	0
50000	20.918	16.801	13.147	0 20000 40000 60000 80000 100000
70000	47.831	38.197	31.588	→ O-Graham → Montuno & Fournier → O-Quickhull
100000	108.946	87.346	70.990	5 G-Granam 5 Montano & Pourmer 5-O-Quicknum

Table 5 The actual running times (in seconds) of the algorithms for sun data.

result on all the data, \mathcal{O} -Quickhull is 1.431 times faster than \mathcal{O} -Graham algorithm and 1.401 times faster than Montuno and Fournier's algorithm. The reason is that after finding an extreme point of $\mathrm{COCH}(P)$, a large 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 Proposition 3(ii) and Remark 4). Furthermore, \mathcal{O} -

		Montuno and		100
Input	O-	Fournier 's	O-	
	Graham	algorithm	Quickhull	90
1000	0.156	0.347	0.137	80
3000	0.251	0.339	0.177	
5000	0.311	0.451	0.240	70
7000	0.442	0.681	0.327	60 ///
10000	0.891	0.867	0.653	
13000	1.158	1.134	0.869	50
15000	1.811	1.412	1.112	40
17000	2.026	1.725	1.407	
20000	2.834	2.212	2.016	30
25000	4.099	5.118	3.567	20
30000	5.887	5.791	4.334	
35000	8.163	6.440	5.241	10
40000	11.558	7.591	7.032	0
45000	14.476	11.996	10.496	0 20000 40000 60000 80000 100000
50000	17.234	14.745	13.709	
70000	40.323	31.717	29.320	→O-Graham → Montuno & Fournier →O-Quickhull
100000	89.321	76.787	73.227	

Table 6 The actual running times (in seconds) of the algorithms for hollow sun data.

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.

Data types	The ratio of $\mathcal{O} ext{-}\mathrm{Graham}$ to $\mathcal{O} ext{-}\mathrm{Quickhull}$	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

QUICKHULL does not need a preprocessing step, which rearranges the input points as \mathcal{O} -graham algorithm and Montuno and Fournier's algorithm.

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]) can be given. In addition, we can use the idea of space subdivision [16] to give efficient algorithms for finding such hulls. They will be the subject of another paper.

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Appendix

The proof of Proposition 2

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. 20 (i)).

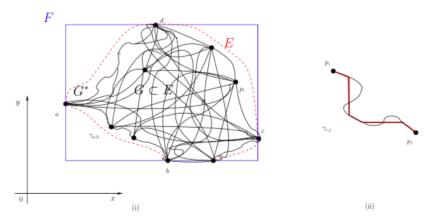


Fig. 20 (i) $G:=\bigcup_{(i,j)\in[1,m]\times[1,m]}\gamma_{i,j}$ and the region G^* formed by the path γ . (ii) Adapting $\gamma_{i,j}$ to get an orthogonal convex set.

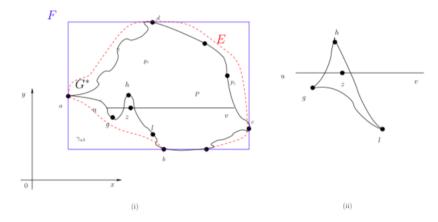


Fig. 21 $z \in [u, v]$ in the region G^* formed by the monotone paths $\gamma_{h,g}, \gamma_{h,l}$ and $\gamma_{g,l}$.

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. 20 (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. 21 (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. 21 (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.

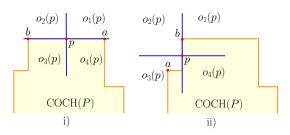
The proof of Lemma 4

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. 22) and p is in the interior of COCH(P) (see Fig. 23).

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 p and the remaining two orthants contain the extreme point p (see Figure 22 i)). As two some orthants of p contain the extreme point p, without loss of generality assume that p belongs to the rectilinear half line containing b (see Fig. 22 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. 22 ii)) contain at least one other extreme point of $\mathrm{COCH}(P)$. Indeed, suppose that $o_4(p)$ does not contain any extreme point of $\mathrm{COCH}(P)$. Then, by Lemma 1(ii), $o_4(p)$ does not contain any point of P. Let c be a smallest y-coordinate point among the points of P in $o_1(p)$, 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 $\mathrm{l}(c,d)$ is an \mathcal{O} -support and the intersection of $\mathrm{l}(c,d)$ and $\mathrm{l}(a,b)$ consists of two distinct points. Therefore, the 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).



 $o_2(p)$ $o_1(p)$ $o_3(p)$ $o_4(p)$ $o_4(p)$

Fig. 22 p lies on the boundary of COCH(P).

Fig. 23 p is in the interior of COCH(P).

The proof of Lemma 5

Proof. Let $p \in \mathcal{M}(P)$, there is an its orthant that 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 that $o_2(p)$ does not contain any points of $\mathrm{COCH}(P) \setminus \{p\}$. We will prove this by contradiction. Indeed, suppose that there exists $u \in o_2(p) \cap (\mathrm{COCH}(P) \setminus \{p\})$. There are two following cases:

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

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

i.e., $p \in \text{o-ext}(\text{COCH}(P))$. Therefore

$$\mathcal{M}(P) \subseteq \text{o-ext}(\text{COCH}(P)).$$
 (4)

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(i) implies $p \in P$. According to Definition 4, there exists an orthant o(p) of p that does not contain any point of $\text{COCH}(P) \setminus \{p\}$. We can conclude from Lemma 1(ii) that o(p) also does not contain any points of $P \setminus \{p\}$, i.e., $p \in \mathcal{M}(P)$. Therefore

$$o\text{-ext}(COCH(P)) \subseteq \mathcal{M}(P).$$
 (5)

It follows from (4) and (5) that
$$\mathcal{M}(P) = \text{o-ext}(\text{COCH}(P))$$
.

The proof of Theorem 1

Proof. If $|P_{ab}| = 0$, then the output of \mathcal{O} -Quickhull has no points, which is the satisfied conclusion.

If $|P_{ab}| \geq 1$, the function \mathcal{O} -Quickhull (a, b, P_{ab}) of \mathcal{O} -Quickhull chooses the farthest point c from $\mathcal{L}^v(a, b)$ and therefore c is an extreme point with index j of COCH(P) (Proposition 3(i)).

$$t$$
 is in the rectangle with the diagonal $[c_1, v_1]$. (6)

We prove on the contrary that if $c_1 \in P_{ab}, c_1 \notin \{a, b\}$ is an extreme point with index j of COCH(P) then c_1 is also generated by some step in \mathcal{O} -QUICKHULL, that is, there exist a_1, b_1 in Case j are two distinct extreme points with index j of COCH(P) such that c_1 is the farthest point from the directed orthogonal line $\mathcal{L}^{v_1}(a_1, b_1)$. Indeed, a_1 (b_1 , resp.) can be chosen as the extreme point with the smallest x-coordinate (largest y-coordinate, resp.) of the extreme points with the same index j but its x-coordinate (y-coordinate,

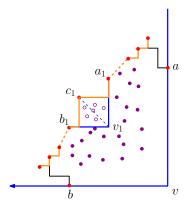


Fig. 24 Three orthogonal extreme points a_1, c_1, b_1 with in dex 2.

resp.) is greater (less, resp.) than that of c_1 . Consider Case j=2 (the other cases are similar). Take a point $t \in P_{a_1b_1}$, we claim that

Assume the contrary that t is outside the rectangle with the diagonal $[c_1, v_1]$, i.e., t in $o_2(c_1)$ or $P_{a_1c_1}$ or $P_{c_1b_1}$. If $t \in o_2(c_1)$ then contradicts the assumption that c_1 is an extreme point with index 2. If $t \in P_{a_1c_1}$ (or $t \in P_{c_1b_1}$, resp.), i.e., $P_{a_1c_1} \neq \emptyset$ (or $P_{c_1b_1} \neq \emptyset$, resp.), then there exists a farthest point u of $P_{a_1c_1}$ (or $P_{c_1b_1}$, resp.) from the directed orthogonal line from a_1 to c_1 (or from c_1 to b_1 , resp.). According to Proposition 3(i), u is an extreme point with index 2 of $P_{a_1c_1}$ (or $P_{c_1b_1}$, resp.). It follows that $x_{c_1} < x_u < x_{a_1}$ (or $y_{b_1} < y_u < y_{c_1}$), resp.). This contradicts the choice of a_1 and b_1 . Thus (6) holds true. It follows that $dist(t, v_1) \leq dist(c_1, v_1)$ and therefore c_1 is the farthest point from the directed orthogonal line $\mathcal{L}^{v_1}(a_1, b_1)$.