# Convex Hull Algorithms: Jarvis's March and Graham's Scan

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## Convexity

Introduce the rubberband.

A subset  $\mathcal{C} \subseteq \mathbb{R}^2$  is called *convex* if and only if for all  $p, q \in S$ ,  $(1 - \lambda)p + \lambda q \in \mathcal{C}$  for all  $\lambda \in [0, 1]$ . Other equivalent definitions—

- A set  $\mathcal{C} \subseteq \mathbb{R}^2$  is convex if and only if it can be expressed as the intersection of (possible infinitely many) closed half-spaces.
- A set  $\mathcal{C} \subseteq \mathbb{R}^2$  is convex if and only if for every point  $p \in \mathbb{R}^2 \setminus \mathcal{C}$ , there exists a hyper plane diving the space into one open half-space containing p and one closed half-space containing  $\mathcal{C}$ .

Definition can be easily extended to higher dimensions.

## Convex Hull

The convex hull  $\mathcal{CH}(S)$  of a set  $S \subseteq \mathbb{R}^2$  is the 'smallest' convex set that contains S. More precisely,

• The convex hull  $\mathcal{CH}(S)$  of a set S is the intersection of all convex sets that contain S:

$$\mathcal{CH}(S) = \bigcap_{\lambda \in \Lambda} \{ C \subseteq \mathbb{R}^n : S \subseteq C, \ C \text{ is convex} \}.$$

• The convex hull  $\mathcal{CH}(S)$  of a set S consists of all convex combinations of finitely many points in S:

$$\mathcal{CH}(S) = \{ \sum_{i=1}^{k} \lambda_i p_i : p_i \in S, \ \lambda_i \ge 0, \ \sum_{i=1}^{k} \lambda_i = 1, \ k \in \mathbb{N} \}$$

## Vertices

The *vertices*, intuitively, are simply the vertices of the polygon formed by the convex hull. Formally, the *non-vertices* are defined as

$$\overline{V}(S) = \{ p \in S : \mathcal{CH}(S) = \mathcal{CH}(S \setminus \{p\}) \}$$

Thus, the vertices become

$$V(S) = S \backslash \overline{V}(S).$$

### **Problem Statement**

Given a set  $S = \{p_1, p_2, \dots, p_n\}$  of points in  $\mathbb{R}^2$ , compute the ordered set V(S), ordered in clockwise direction.

## Interesting (and Helpful) Results

First introduce Jarvis and Graham; they developed fantastic convex hull algorithms for this problem statement.

## Caratheodory's Theorem (1907)

If  $p \in \mathcal{CH}(S)$  for  $S \subseteq \mathbb{R}^2$ , then p can be expressed as  $p = \sum_{i=1}^3 \lambda_i p_i$  where  $\lambda_i \geq 0$  and  $\sum_{i=1}^k \lambda_i$  for some  $p_i \in S$ .

All it says is, if p is in the convex hull, then there exists a triangle containing p whose vertices are formed by vertices in S. Gives us an idea that to determine if p is a vertex or not, we need only consider triangles; gave insight to Jarvis that only small subset of poitns is needed to represent a convex hull, and that they can be computed incrementally.

## Radon's Theorem (1921)

Any set of 4 points in  $\mathbb{R}^2$  can be partitioned into two sets whose convex hulls intersect. (Proof is a good exercise). Evoked the idea of divide and conquer.

## Jarvis March (1973)

R. A. Jarvis. Involved in robotics, computer vision, path finding, and image processing.

## Concept (Bruteforce wrapping)

- 1. Start with an empty set  $V(S) = \emptyset$ .
- 2. Consider all  $(p,q) \in S \times S$ .
- 3. If all  $r \in S \setminus \{p,q\}$  exist to the right of the directed line  $\ell(p,q)$ , then set V(S) to be  $V(S) \cup \{p,q\}$ .
- 4. Order clockwise, and return.

### Pseudocode

#### **Algorithm 1** Jarvis's March (Bruteforce Wrapping)

```
Require: A finite set of points S \subset \mathbb{R}^2.
Ensure: The set of extreme points V(S), arranged in a clockwise order.
 1: Initialize V(S) \leftarrow \emptyset.
 2: for all distinct pairs (p,q) \in S \times S with p \neq q do
        Assume \ell(p,q), the directed line through p and q, is a boundary of
    V(S).
 4:
        for all points r \in S \setminus \{p, q\} do
            if r lies strictly to the left of \ell(p,q) then
 5:
                Discard \ell(p,q) as a boundary.
 6:
            end if
 7:
        end for
 8:
        if \ell(p,q) was never discarded then
 9:
            Include p and q in V(S).
10:
11:
        end if
12: end for
13: Order the elements of V(S) in a clockwise manner.
14: Return V(S).
```

## Time Complexity

$$O(n^2h) + O(h \log h) = O(n^2h), \quad h = \text{number of vertices}$$

Really, on the worst case possible when every point in S is a vertex, this is  $O(n^3)$ .

Revolutionary, but completely unoptimized.

Gift wrapping version is much better; the concept goes as follows:

## Concept (Gift wrapping)

- 1. Start with an empty set  $V(S) = \emptyset$ .
- 2. Immediately add the 'leftmost' point,  $p_0$  to V(S).
- 3. You now only need to consider  $(p_0, p) \in S \times S$  where  $p \neq p_0$ .

- 4. If all  $r \in S \setminus \{p_0, p\}$  exist to the right of the directed line  $\ell(p, q)$ , then set  $p_1 = p$  and add  $p_1$  to V(S).
- 5. Continue with  $(p_1, p) \in S \times S$  where  $p \neq p_1$  until you similarly find  $p_2$ . Keep continuing until  $p_h$  is forced to be  $p_0$ . Return.

## **Some Explanations**

### What happens if r lies on $\ell(p,q)$ ?

We could simply add it, increasing the value of h and increasing computation time. But a check for not adding it would preferably; this check would cost us O(h), thus not increasing computation time by much.

### How to check if r lies to the right of $\ell(p,q)$ ?

You could find the angle between  $\ell(p,q)$  and  $\ell(q,r)$ . A better approach is to compute

$$D = \det \begin{pmatrix} 1 & p_x & p_y \\ 1 & q_x & q_y \\ 1 & r_x & r_y \end{pmatrix}$$

The sign of D tells whether r lies to the right (and on) or the left of the line.