

Geometric Algorithms – The Convex Hull Problem in 2 & 3 Dimensions

Rehman, M. Usaid Ali, Syed Anus Ozair, Faraz

Habib University

August 21, 2021



Outline

- 1 Introduction to Computational Geometry
 - Overview
 - History & Famous Problems
- 2 Geometric Preliminaries
 - Definitions & Notation
- 3 The Convex Hull Problem
 - Introduction & Intuition
 - Problem Definition
- 4 Convex Hull Algorithms & Complexity
 - Lower Bound & Output-Sensitivity
 - Algorithms
- 5 3D-Convex Hulls & More



Outline

1 Introduction to Computational Geometry

■ Overview

■ History & Famous Problems

2 Geometric Preliminaries

■ Definitions & Notation

3 The Convex Hull Problem

■ Introduction & Intuition

■ Problem Definition

4 Convex Hull Algorithms & Complexity

■ Lower Bound & Output-Sensitivity

■ Algorithms

5 3D-Convex Hulls & More



The Paradigm of Computational Geometry

- A sub-field of algorithm theory involving design and analysis of algorithms with geometric input and output.
- Mostly focuses on the discrete aspect of geometric problem solving.
- Primarily deals with straight or flat objects or simple curves.
- Computational geometry aims to provide basic geometric tools from which application areas can build algorithms.
- It also aims to provides theoretical analytical tools to analyze the performance of these algorithms.



Outline

1 Introduction to Computational Geometry

- Overview

- History & Famous Problems

2 Geometric Preliminaries

- Definitions & Notation

3 The Convex Hull Problem

- Introduction & Intuition

- Problem Definition

4 Convex Hull Algorithms & Complexity

- Lower Bound & Output-Sensitivity

- Algorithms

5 3D-Convex Hulls & More



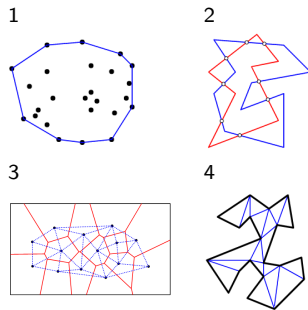
A Brief History & Timeline

- Geometry has been central to mathematics since the time of the Ancient Greeks and the Ancient Egyptians.
- Euclid's *Elements* had a profound influence on the study of geometry.
- Introduction of coordinates permitted an increase in computational power.
- The term "Computational Geometry" was coined first by Marvin Minsky in his book "Perceptrons" in 1969.
- Multiple application areas have been the incubation bed for this discipline.
- These problems include Euclidean traveling salesman, minimum spanning tree, linear programming etc.



Famous Problems in Computational Geometry

- 1 Convex Hulls
- 2 Intersections
- 3 Voronoi Diagrams and Delaunay Triangulations
- 4 Triangulation and Partitioning
- 5 Optimization and Linear Programming
- 6 Geometric Search Problems
 - Nearest-neighbor searching
 - Point location



Outline

- 1 Introduction to Computational Geometry
 - Overview
 - History & Famous Problems
- 2 Geometric Preliminaries
 - Definitions & Notation
- 3 The Convex Hull Problem
 - Introduction & Intuition
 - Problem Definition
- 4 Convex Hull Algorithms & Complexity
 - Lower Bound & Output-Sensitivity
 - Algorithms
- 5 3D-Convex Hulls & More



Basics of Euclidean Geometry

- The main objects considered in computational geometry are sets of points in Euclidean space.
- Sets of points are finite, or at least *finitely specifiable*.
- By \mathbb{R}^d , we denote the *d-dimensional Euclidean space*.
- A *d*-tuple (x_1, \dots, x_d) denotes a point of \mathbb{R}^d where $x_i \in \mathbb{R}$.
- A *polygon* in \mathbb{R}^2 is defined by a finite set of line segments where each extreme point is shared by exactly two segments called *edges*.
- A *polyhedron* in \mathbb{R}^3 is defined by a finite set of plane polygons such that every edge of a polygon is shared by only one other polygon.



Outline

- 1 Introduction to Computational Geometry
 - Overview
 - History & Famous Problems
- 2 Geometric Preliminaries
 - Definitions & Notation
- 3 The Convex Hull Problem
 - Introduction & Intuition
 - Problem Definition
- 4 Convex Hull Algorithms & Complexity
 - Lower Bound & Output-Sensitivity
 - Algorithms
- 5 3D-Convex Hulls & More



What is a Convex Hull?

- A *convex set* is a subset of Euclidean space where given any two points in the set, the set contains the whole line segment joining them.
- A *convex hull* of a shape is the smallest convex set containing the shape.
- The convex hull can also be defined as the intersection of all convex sets of a given subset of Euclidean space.

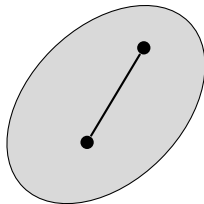


Figure: A convex set

Mixing Things

- Imagine you have the following mixtures of substances:

subst.	fract. A	fract. B
s_1	10%	35%
s_2	20%	5%
s_3	40%	55%

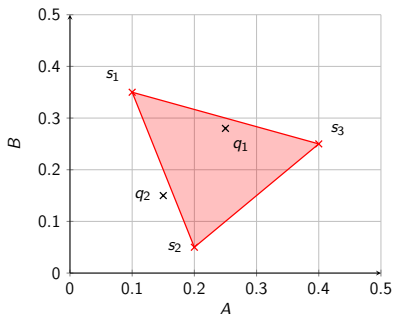
- Using these substances can we create two other substances s_{q_1} and s_{q_2} ?

q_1	25%	28%
q_2	15%	15%



Mixing Things

- Let's plot these values:



- We can observe that a substance $q \in \mathbb{R}^2$ can be created using substances in $S \subset \mathbb{R}^2$ if and only if $q \in \text{CH}(S)$.



Outline

- 1 Introduction to Computational Geometry
 - Overview
 - History & Famous Problems
- 2 Geometric Preliminaries
 - Definitions & Notation
- 3 The Convex Hull Problem
 - Introduction & Intuition
 - Problem Definition
- 4 Convex Hull Algorithms & Complexity
 - Lower Bound & Output-Sensitivity
 - Algorithms
- 5 3D-Convex Hulls & More



What is the Convex Hull Problem?

- The convex hull problem is formally defined as:

Given a set S of N points in \mathbb{R}^d , construct its convex hull $\text{CH}(S)$.

- Another version of the convex hull problem is:

Given a set S of N points in \mathbb{R}^d , identify those that are vertices of $\text{conv}(S)$.

- Both versions of the problem have the same asymptotic hardness since one problem can be transformed to the other.



Convex Hull for Finite Set of Points

- The convex hull for a finite set of points in the plane produces a convex polygon.

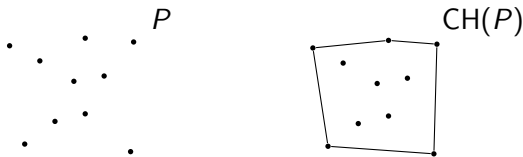


Figure: A point set and its convex hull.

- In 3 dimensions however, the convex hull is not a convex polygon, but instead a convex polyhedron – generalizes to a polytope in higher dimensions.



Outline

1 Introduction to Computational Geometry

- Overview
- History & Famous Problems

2 Geometric Preliminaries

- Definitions & Notation

3 The Convex Hull Problem

- Introduction & Intuition
- Problem Definition

4 Convex Hull Algorithms & Complexity

- Lower Bound & Output-Sensitivity
- Algorithms

5 3D-Convex Hulls & More



Lower Bound on Computational Complexity

Overview and Approach

- The convex hull problem outputs a polygon – a cyclic enumeration of the boundary vertices.
- We can say that we have to produce a sorting of the vertices. which has a lower bound of $\Omega(n \log n)$.
- To find lower bound of convex hull problem, we will reduce the sorting problem to the convex hull problem in $O(n)$ time, which implies:

Theorem

Any algorithm for the convex hull problem requires $\Omega(n \log n)$ time in the worst case.



Lower Bound on Computational Complexity

Reduction of Sorting to Convex Hull

- Let $X = \{x_1, \dots, x_n\}$ be the n values we wish to sort.
- We can "lift" each of these points onto a parabola $y = x^2$, by mapping x_i to the point $p_i = (x_i, x_i^2)$.
- P is the resulting set of points – all of which lie on the convex hull.
- The sorted order of points along the lower hull is the same as the sorted order X .

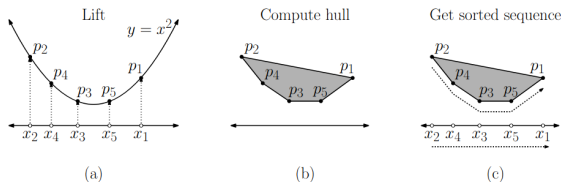


Figure: Reduction from sorting to convex hull.



Output-Sensitive Algorithms

- Some convex hull algorithms depend both on n input points, and h output points.
- Such algorithms are called *output-sensitive* algorithms.
- They may be asymptotically more efficient than $\Theta(n \log n)$ algorithms when $h = o(n)$.
- The lower bound for these algorithms is $\Omega(n \log h)$ in the planar case.
- Kirkpatrick and Seidel devised an algorithm that achieved this in 1986.
- In 1996, a much simpler algorithm was devised by Chan called Chan's algorithm.



Outline

- 1 Introduction to Computational Geometry
 - Overview
 - History & Famous Problems
- 2 Geometric Preliminaries
 - Definitions & Notation
- 3 The Convex Hull Problem
 - Introduction & Intuition
 - Problem Definition
- 4 Convex Hull Algorithms & Complexity
 - Lower Bound & Output-Sensitivity
 - Algorithms
- 5 3D-Convex Hulls & More



The General Position Assumption

- Before we look at algorithms, let's look at an important assumption.
- We assume that the points are in *general position*.
- This means that degenerate configurations do not arise in input.
- Point sets in general position maintain general position if perturbed infinitesimally.
- Merely a convenience to avoid dealing with a lot of special cases during algorithm design.



Convex Hull Algorithms

- There are multiple algorithms that employ different techniques to compute the convex hull of a point set.
- Following are some of these algorithms with their time complexity:

Algorithm	Time Complexity
Jarvis March	$O(nh)$
Graham Scan	$O(n \log n)$
Quickhull	$O(n \log n)$
Andrew's Algorithm	$O(n \log n)$
Kirkpatrick-Seidel Algorithm	$O(n \log h)$
Chan's Algorithm	$O(n \log h)$

Table: Some algorithms and their complexities.



Jarvis March

Algorithm Description & Visualization

- Jarvis march a.k.a the gift wrapping algorithm, was published in 1973 by R. A. Jarvis.
- The algorithm begins with $i \leftarrow 0$, and a point p_0 known to be on the convex hull.
- Then it selects the point p_{i+1} such that all points are to the right of the line segment $p_i p_{i+1}$.
- This can be done in $O(n)$ time by comparing polar angles.
- Let $i \leftarrow i + 1$, and repeat until algorithm $p_h \leftarrow p_0$ again.
- This would yield the convex hull in h steps.
- [Here is a link to an animation for Jarvis march.](#)



Jarvis March

Performance & Time Complexity

- Jarvis march is among the least efficient algorithms for convex hull.
- The algorithm loops h times and within each iteration, it loops n times.
- Therefore, the time complexity is $O(nh)$.

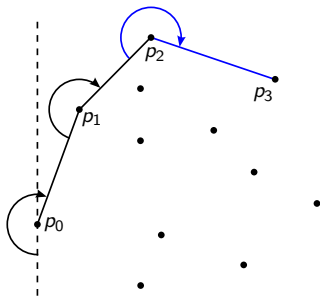


Figure: The Jarvis march procedure.



Graham Scan

Algorithm Description & Visualization

- Based on the *incremental construction* technique.
- First, select the lowest point in the point set P called point p_0 .
- Sort all points cyclically w.r.t point p_0 based on polar angles.
- For each point, determine if travelling to that point constitutes a left turn or a right turn.
- If right turn: second-to-last point is not part of $CH(P)$ but lies inside.
- Do the same for previously visited points until 'left-turn set' is encountered.
- Repeat until starting point is reached in a counter-clockwise fashion.
- [Link to visualization.](#)



Graham Scan

Pseudocode

Algorithm 1: GRAHAM SCAN

Input: A point set P **Output:** The convex hull of P , $CH(P)$

- 1 let p_0 be the minimum and left-most point in P
 - 2 let $\langle p_1, p_2, \dots, p_n \rangle$ be the remaining points in P sorted by polar angle in counterclockwise order around p_0
 - 3 let S be an empty stack
 - 4 PUSH(p_0, S)
 - 5 PUSH(p_1, S)
 - 6 PUSH (p_2, S)
 - 7 for $i \leftarrow 3$ to $i \leftarrow n$ do
 - 8 while the angle formed by points TOP(S) NEXT_TO_TOP(S),
 and p_i make a non-left turn do
 - 9 POP(S)
 - 10 PUSH(p_i, S)
 - 11 return S
-



Graham Scan

Performance & Time Complexity

- Graham Scan is an efficient algorithm compared to Jarvis march.
- The sorting procedure is done in $O(n \log n)$ time.
- The loop is $O(n)$ and not $O(n^2)$.
- We visit each point at most twice – once as a left turn and once as a right turn.
- The time to sort dominates the time complexity of the algorithm.
- Therefore, Graham scan is $O(n \log n)$.



Quickhull Algorithm

The Philosophy Behind the Algorithm

- The Quickhull algorithm is a divide-and-conquer algorithm.
- It is based on the strategy of triangular expansion.
- It is also used for computing convex hull in higher dimensions due to its efficiency.
- Quickhull also uses the prune-and-search approach, where the input size is reduced by a constant factor at each step.
- Its working can be visualized [here](#).



Quickhull Algorithm

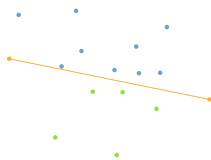
Pseudocode for 2-Dimensional Quickhull

- 1 Find the points with minimum and maximum x -coordinates.
- 2 Use the line formed by the two points to divide the set into two subsets of points.
- 3 Determine the point, on one side of the line, with the maximum distance from the line — forms a triangle with those of the line.
- 4 The points lying inside of that triangle cannot be part of the convex hull and can be ignored in later steps.
- 5 Repeat the previous two steps on the lines formed by the triangle (recursion).
- 6 Keep on doing so until no points are left. The points selected constitute the convex hull.

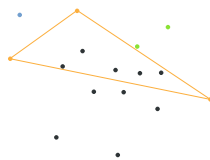


Quickhull Algorithm

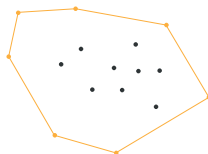
Visualization of Algorithm



(a) Steps 1 – 2



(b) Steps 3 – 5



(c) Step 6

Quickhull Algorithm

Performance & Time Complexity

- Finding the extreme points is an $O(n)$ operation.
- Quickhull divides into two sub-problems.
- In the best case, the partition size is well-balanced:

$$T(n) = 2T(n/2) + O(n)$$

$$T(n) = O(n \log n)$$

- Whereas, in the worst case this degenerates to $T(n) = nT(n-1) + cn$, which means the worst case is quadratic.



The 3D Convex Hull Problem

- Recall that in 3 dimensions, the convex hull is a polyhedron (3-polytope).
- In this case, we are concerned with the *facets* of a polytope.
- The convex polytope is the convex hull of all its vertices.
- More complicated problem which is harder to visualize and implement.



Figure: A 3-dimensional convex hull.

The Gift-Wrapping Algorithm

- The most famous and the simplest algorithm is the gift wrapping algorithm.
- It is the extension of the 2-dimensional Jarvis march algorithm.
- The steps of the algorithm are:
 - 1 Find a facet that is guaranteed to be on the convex hull.
 - 2 Repeat:
 - 1 Find an edge e of a face f that's on the CH, and such that the face on the other side of e has not been found.
 - 2 For all remaining points π , find the angle (e, π) with f .
 - 3 Find point π with the minimal angle; add face (e, π) to CH.
- The algorithm runs in $O(n \cdot F)$ time where F is the number of faces on the convex hull.



Other Approaches

- Approaches regarding 3-dimensional convex hulls mostly extend to N dimensions.
- The N -dimensional convex hull is an N -polytope.
- The gift-wrapping algorithm extends to N -dimensions.
- There are also other methods such as:
 - N -dimensional Quickhull
 - Chan's Algorithm
 - Divide-and-Conquer
 - Beneath-and-Beyond Method



References



M. I. Shamos & F. P. Preparata

Computational Geometry: An Introduction.

Springer New York, 1985.



M. de Berg (ed.)

Computational Geometry: Algorithms and Applications.

Springer, 2010



T.H. Cormen, C.E. Leiserson, R. Rivest, C. Stein

Introduction to Algorithms

MIT Press, 1989

