

# Convex Hull

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January 5, 2026

# Contents

- ① Introduction
- ② Concept Overview
- ③ Rubber Band Intuition
- ④ Orientation Test
- ⑤ Naive Algorithm
- ⑥ Graham Scan
- ⑦ Jarvis March
- ⑧ Divide and Conquer
- ⑨ QuickHull
- ⑩ Comparison
- ⑪ Applications
- ⑫ Conclusion

# Introduction

Convex Hull is a fundamental problem in **computational geometry**.

**Definition:** The convex hull of a set of points is the smallest convex polygon that encloses all points.

## Applications:

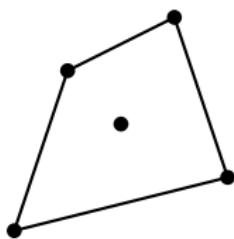
- Computer Graphics
- Collision Detection
- GIS

# Concept Overview

Given a set of points in a 2D plane:

- Convex hull connects the outermost points
- All interior points lie inside the polygon
- Boundary points are called extreme points

# Rubber Band Analogy



Imagine stretching a rubber band around the points. When released, it forms the convex hull.

# Orientation Test

For points  $A(x_1, y_1)$ ,  $B(x_2, y_2)$ ,  $C(x_3, y_3)$ :

$$val = (x_2 - x_1)(y_3 - y_1) - (y_2 - y_1)(x_3 - x_1)$$

## Interpretation:

- $val = 0 \rightarrow$  Collinear
- $val > 0 \rightarrow$  Counter-Clockwise
- $val < 0 \rightarrow$  Clockwise

## Orientation Test – Worked Example

Let:

$$A(1, 1), B(3, 1), C(2, 2)$$

Compute:

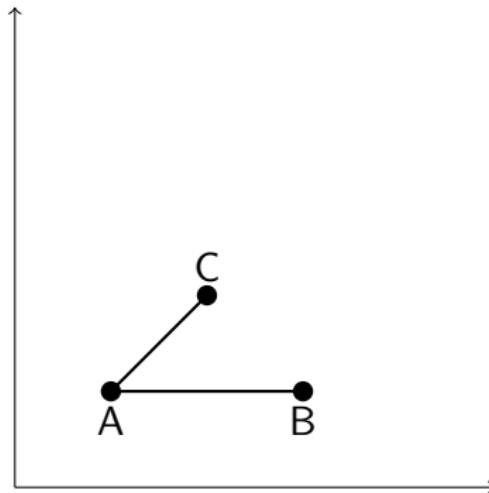
$$val = (x_B - x_A)(y_C - y_A) - (y_B - y_A)(x_C - x_A)$$

$$= (3 - 1)(2 - 1) - (1 - 1)(2 - 1)$$

$$= 2 \times 1 - 0 = 2$$

Since  $val > 0$ , the orientation is **Counter-Clockwise (Left Turn)**.

# Orientation Test – Diagram



Since point C lies to the left of line AB, the orientation is **Counter-Clockwise**.

# Naive (Brute Force) Algorithm

## Idea:

- For every pair of points ( $P_i, P_j$ )
- Check if all other points lie on one side
- If yes  $\rightarrow$  hull edge

**Time Complexity:**  $O(n^3)$  **Space Complexity:**  $O(1)$

# Naive Algorithm – Numerical Example

Consider edge candidate:

$$P_1(1, 1), P_2(3, 1)$$

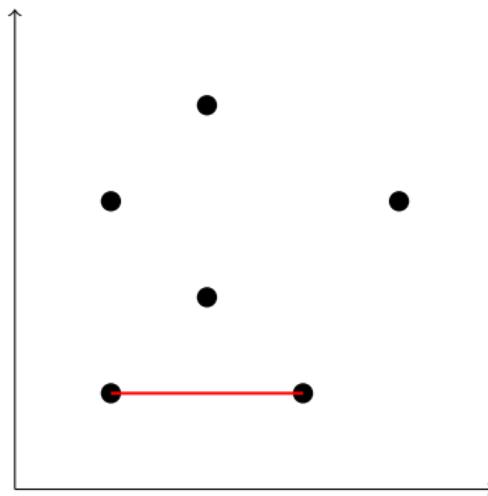
Check remaining points:

- $(2, 2) \rightarrow$  above the line
- $(4, 3) \rightarrow$  above the line
- $(2, 4) \rightarrow$  above the line
- $(1, 3) \rightarrow$  above the line

All points lie on the **same side**.

**Conclusion:** Edge  $(1, 1) - (3, 1)$  belongs to the convex hull.

## Naive Algorithm – Edge Verification



All other points lie above the red line.

**Conclusion:** Edge  $(1, 1) - (3, 1)$  is part of the convex hull.

# Naive Convex Hull – Time Complexity Analysis

Let  $n$  be the number of input points.

## Step 1: Selecting Point Pairs

- Consider every pair of points  $(P_i, P_j)$  as a candidate hull edge
- Total number of pairs:  $\binom{n}{2} = O(n^2)$

## Step 2: Checking Side of Line

- For each pair  $(P_i, P_j)$ , check all remaining points
- Use orientation / line equation to determine the side
- Number of remaining points:  $n - 2 \approx O(n)$
- Time per pair:  $O(n)$

## Step 3: Hull Edge Decision

- If all points lie on the same side, the edge is part of the convex hull
- Otherwise, discard the edge

## Total Time Complexity

$$T(n) = O(n^2) \times O(n) = \boxed{O(n^3)}$$

## Space Complexity

$$\boxed{O(1)}$$

(Only constant extra variables are used)

# Graham Scan – Core Idea

Graham Scan constructs the convex hull by:

- Sorting points by polar angle
- Removing points that cause right turns
- Using a stack to maintain convexity

# Graham Scan – Algorithm Steps

- ① Find the point with lowest y-coordinate (pivot  $P_0$ )
- ② Sort all other points by polar angle with respect to  $P_0$
- ③ Push first three points into a stack
- ④ For each remaining point:
  - While the top two stack points and current point make a clockwise turn, pop the stack
  - Push the current point
- ⑤ Stack contains convex hull

# Graham Scan – Example (Step 1 & 2)

Given points:

$$(1, 1), (2, 2), (3, 1), (4, 3), (2, 4), (1, 3)$$

## Step 1: Pivot Selection

Lowest y-coordinate point:

$$P_0 = (1, 1)$$

## Step 2: Polar Angle Sorting

Sorted order:

$$(1, 1), (3, 1), (4, 3), (2, 4), (1, 3), (2, 2)$$

# Graham Scan – Example (Stack Operations)

Initial stack:

$$(1, 1), (3, 1), (4, 3)$$

Check next point (2, 4):

$$\text{Orientation}((3, 1), (4, 3), (2, 4)) = \text{CCW}$$

Push (2, 4)

Next point (2, 2):

$$\text{Orientation}((4, 3), (2, 4), (2, 2)) = \text{CW}$$

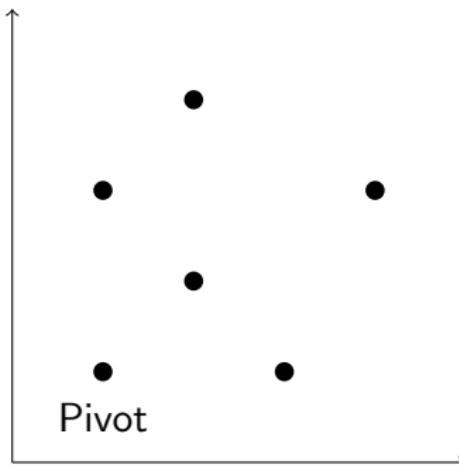
Pop (2, 4)

**Final Hull Points:**

$$(1, 1), (3, 1), (4, 3), (2, 4), (1, 3)$$

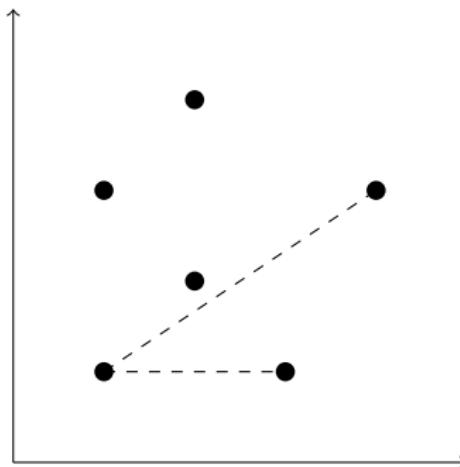
# Graham Scan – Code and Geometry Sync

**Step 1:** Select Pivot  $P_0$



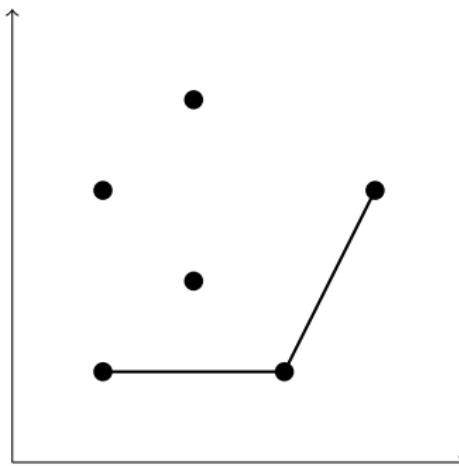
# Graham Scan – Code and Geometry Sync

**Step 2:** Sort by polar angle



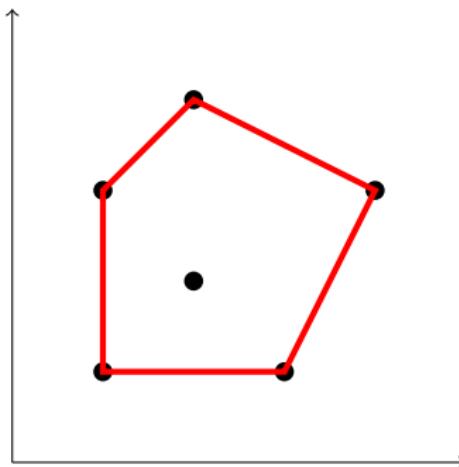
# Graham Scan – Code and Geometry Sync

**Step 3:** Push points to stack



# Graham Scan – Code and Geometry Sync

**Step 4:** Remove clockwise turns



# Graham Scan – Time Complexity Analysis

Let  $n$  be the number of input points.

## Step 1: Finding the Pivot Point

- Scan all points to find the lowest y-coordinate
- Time Complexity:  $O(n)$

## Step 2: Sorting by Polar Angle

- Sort  $n - 1$  points based on polar angle w.r.t pivot
- Each comparison uses orientation test ( $O(1)$ )
- Time Complexity:  $O(n \log n)$

## Step 3: Stack Initialization

- Push first three points into stack
- Time Complexity:  $O(1)$

## Step 4: Stack Processing (Scan Phase)

- Each point is pushed once and popped at most once
- Total stack operations  $\leq 2n$
- Time Complexity:  $O(n)$

## Total Time Complexity

$$T(n) = O(n) + O(n \log n) + O(1) + O(n) = \boxed{O(n \log n)}$$

# Jarvis March – Intuition

Jarvis March works like wrapping a gift:

- Start from an extreme point
- Repeatedly select the most counter-clockwise point
- Stop when the start point is reached

# Jarvis March – Algorithm

- ① Find the leftmost point
- ② Set it as the current hull point
- ③ For every other point, find the one that makes the most CCW turn
- ④ Add it to hull and repeat

# Jarvis March – Worked Example

**Step 1:** Start from leftmost point:

$$(1, 1)$$

**Step 2:** Choose most CCW point:

$$(3, 1)$$

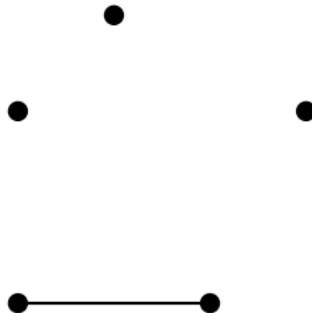
Repeat CCW selection:

$$(3, 1) \rightarrow (4, 3) \rightarrow (2, 4) \rightarrow (1, 3)$$

Return to starting point  $(1, 1)$ .

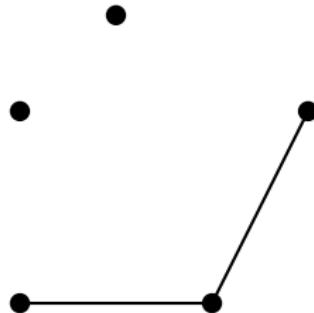
**Hull completed.**

# Jarvis March – Gift Wrapping Animation



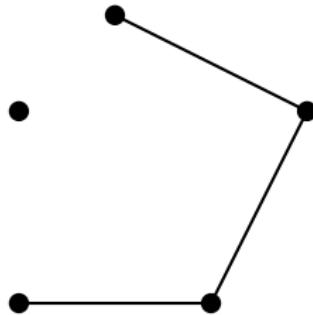
Wrapping proceeds by selecting the most CCW point.

# Jarvis March – Gift Wrapping Animation



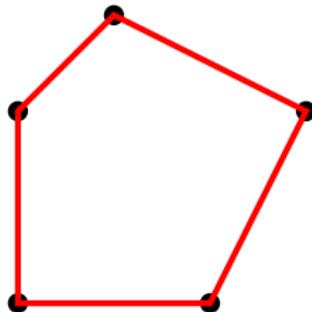
Wrapping proceeds by selecting the most CCW point.

# Jarvis March – Gift Wrapping Animation



Wrapping proceeds by selecting the most CCW point.

# Jarvis March – Gift Wrapping Animation



Wrapping proceeds by selecting the most CCW point.

# Jarvis March (Gift Wrapping) – Time Complexity

Let:

- $n$  = total number of input points
- $h$  = number of points on the convex hull

## Step 1: Find the Starting Point

- Find the leftmost (or lowest) point by scanning all points
- Time Complexity:  $O(n)$

## Step 2: Selecting the Next Hull Point

- For the current hull point, check orientation with all other points
- Select the most counter-clockwise point
- Time Complexity per hull point:  $O(n)$

## Step 3: Wrapping Until Return to Start

- The above step is repeated for each hull point
- Total number of hull points =  $h$

## Total Time Complexity

$$T(n) = O(n) + h \cdot O(n) = \boxed{O(nh)}$$

**Best Case:**  $h$  is small (e.g., triangle)  $\Rightarrow O(n)$

**Worst Case:** All points on hull ( $h = n$ )  $\Rightarrow O(n^2)$

# Divide and Conquer – Strategy

This approach follows the merge sort paradigm:

- Divide the point set
- Solve subproblems recursively
- Merge partial hulls using tangents

# Divide and Conquer – Steps

- ① Sort points by x-coordinate
- ② Divide into left and right halves
- ③ Recursively compute hulls
- ④ Merge hulls using upper and lower tangents

# Divide and Conquer – Example

Sorted by x-coordinate:

(1, 1), (1, 3), (2, 2), (2, 4), (3, 1), (4, 3)

Divide into:

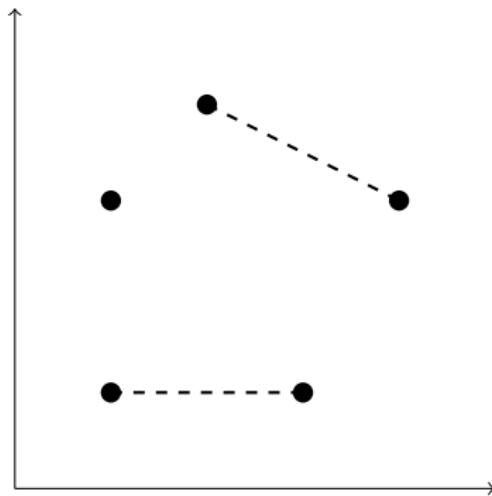
- Left Hull: (1, 1), (1, 3), (2, 4)
- Right Hull: (3, 1), (4, 3)

Merge using upper and lower tangents.

**Final Hull:**

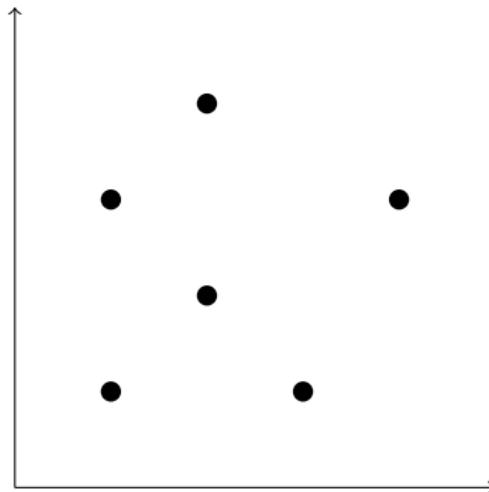
(1, 1), (3, 1), (4, 3), (2, 4), (1, 3)

# Divide and Conquer – Merging Hulls



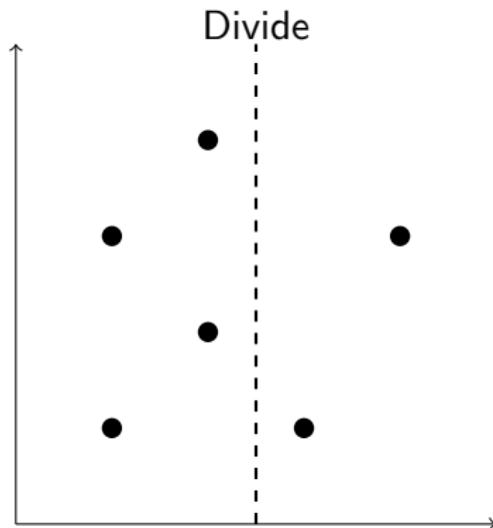
Upper and lower tangents merge the left and right hulls.

## Divide and Conquer – Divide Step



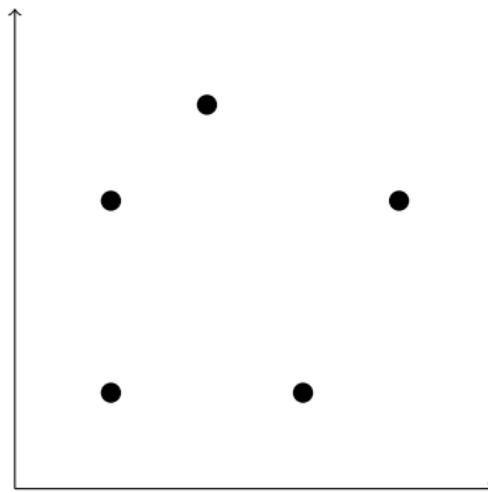
Initial set of points

## Divide and Conquer – Divide Step



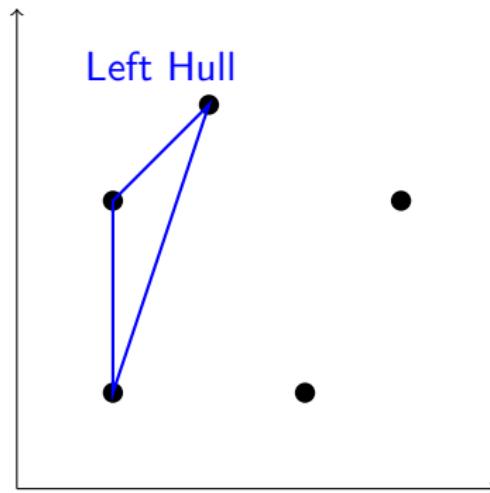
Points divided into left and right halves

# Divide and Conquer – Recursive Hulls



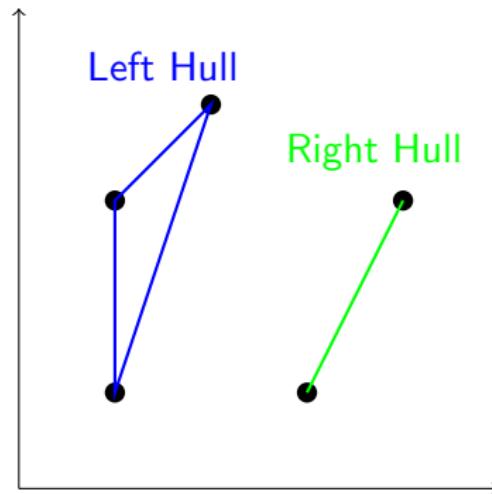
Recursive calls on each half

# Divide and Conquer – Recursive Hulls



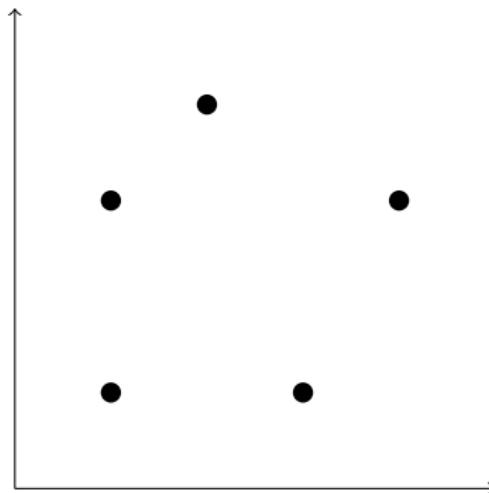
Convex hull of left subset

# Divide and Conquer – Recursive Hulls



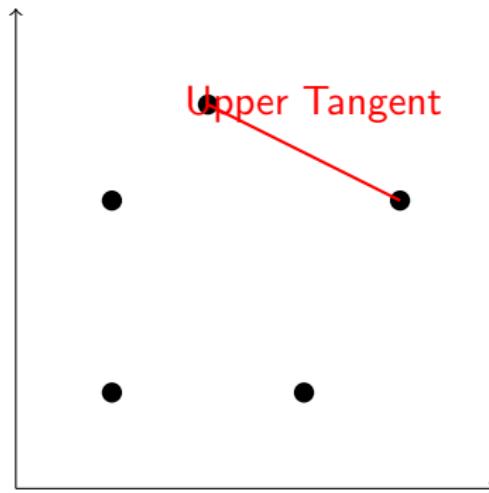
Convex hull of right subset

# Divide and Conquer – Finding Upper Tangent



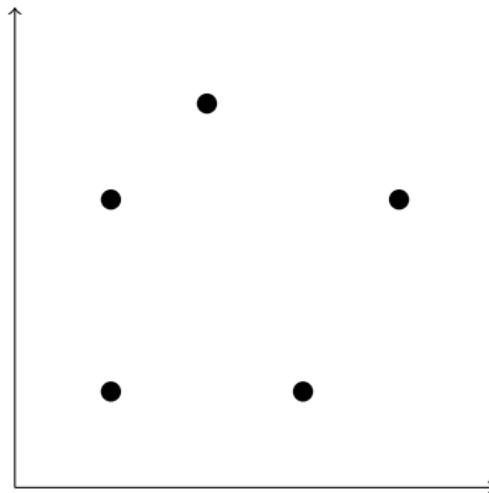
Searching for upper tangent

# Divide and Conquer – Finding Upper Tangent



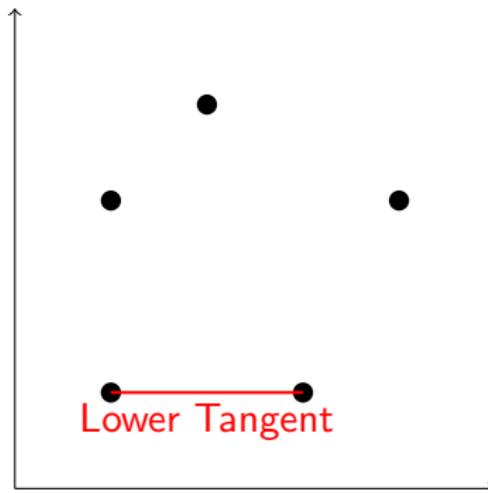
Upper tangent found

# Divide and Conquer – Finding Lower Tangent



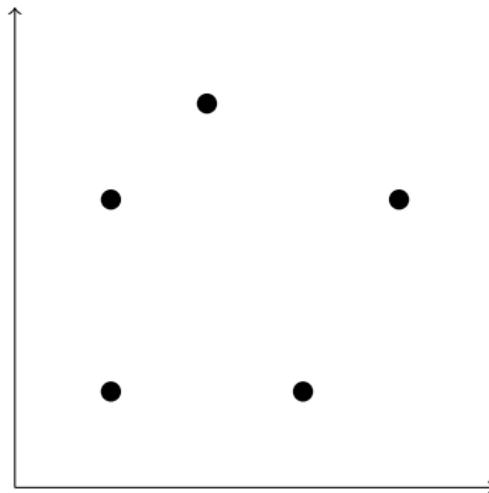
Searching for lower tangent

# Divide and Conquer – Finding Lower Tangent



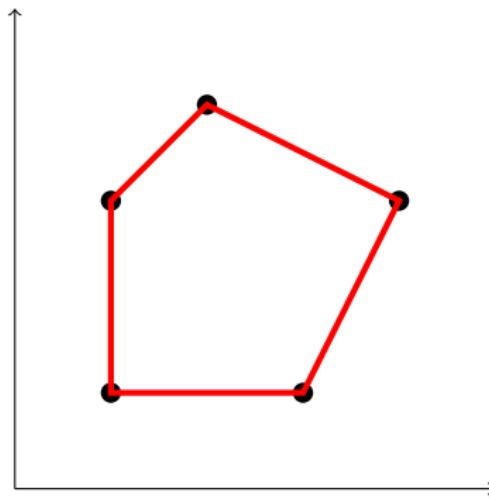
Lower tangent found

# Divide and Conquer – Merge Step



Merging left and right hulls

# Divide and Conquer – Merge Step



Final convex hull

# Divide & Conquer Convex Hull – Recurrence Analysis

Let  $n$  be the number of input points.

## Step 1: Divide

- Sort points by x-coordinate
- Divide the set into two halves of size  $n/2$
- Cost of division (after sorting):  $O(1)$

## Step 2: Conquer

- Recursively compute convex hull for left half
- Recursively compute convex hull for right half
- This gives two subproblems of size  $n/2$

## Step 3: Merge

- Merge the two hulls by finding upper and lower tangents
- Tangent finding and merging takes linear time
- Merge Cost:  $O(n)$

## Recurrence Relation

$$T(n) = 2T\left(\frac{n}{2}\right) + O(n)$$

## Using Master Theorem:

- $a = 2, b = 2, f(n) = O(n)$
- $n^{\log_b a} = n$

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

# QuickHull – Concept

QuickHull is inspired by QuickSort:

- Partition points using a baseline
- Select the farthest point
- Recursively process subsets

# QuickHull – Algorithm Steps

- ① Find leftmost and rightmost points  $A$  and  $B$
- ② Divide points into two sets (above and below  $AB$ )
- ③ Find point  $C$  farthest from  $AB$
- ④ Discard points inside triangle  $ABC$
- ⑤ Recursively apply QuickHull on remaining sets

# QuickHull – Worked Example

**Step 1:** Leftmost and rightmost points:

$$A(1, 1), B(4, 3)$$

**Step 2:** Farthest point from line AB:

$$C(2, 4)$$

Points inside triangle ABC are discarded.

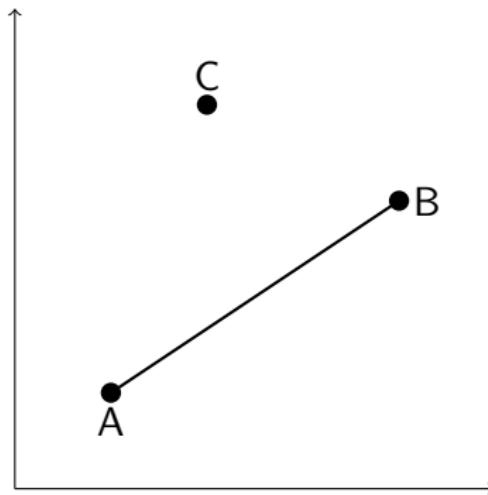
Recursive calls on:

$$A - C \quad \text{and} \quad C - B$$

**Final Hull:**

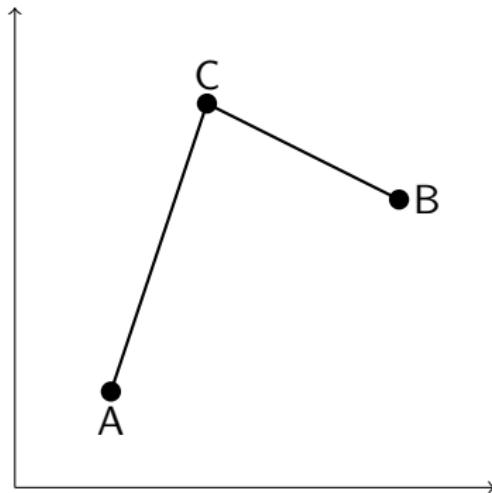
$$(1, 1), (3, 1), (4, 3), (2, 4), (1, 3)$$

# QuickHull – Animated Steps



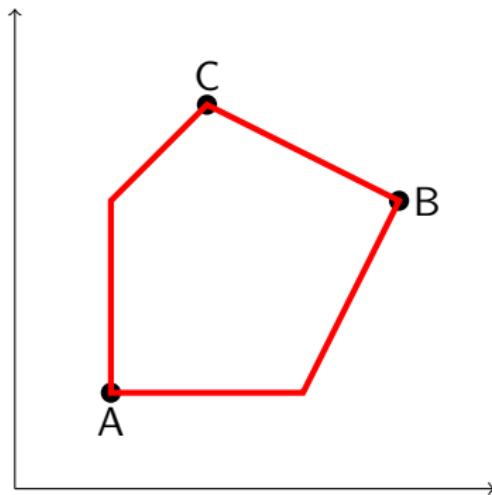
Initial baseline AB

# QuickHull – Animated Steps



Farthest point selected

# QuickHull – Animated Steps



Final convex hull

# QuickHull – Time Complexity Analysis

Let  $n$  be the number of input points.

## Step 1: Initial Partition

- Find leftmost and rightmost points  $A$  and  $B$
- Partition remaining points into two sets (above and below line  $AB$ )
- Time Complexity:  $O(n)$

## Step 2: Farthest Point Selection

- Find the point farthest from the current line
- Distance computed using area (cross product)
- Time Complexity per recursion:  $O(n)$

## Step 3: Recursive Partitioning

- Recursively apply QuickHull on subsets formed by lines  $AC$  and  $CB$
- Points inside the triangle are discarded

## Average Case Analysis

- Points are fairly evenly distributed
- Each recursion splits the set roughly in half

$$T(n) = T\left(\frac{n}{2}\right) + T\left(\frac{n}{2}\right) + O(n) \Rightarrow O(n \log n)$$

## Worst Case Analysis

- All points lie on the convex hull
- Only one point is eliminated per recursion

$$T(n) = T(n - 1) + O(n) \Rightarrow O(n^2)$$

# Algorithm Comparison

Algorithm	Approach	Time Complexity	Space
Naive (Brute Force)	Edge checking	$O(n^3)$	$O(1)$
Graham Scan	Sorting + Stack	$O(n \log n)$	$O(n)$
Jarvis March	Gift Wrapping	$O(nh)$	$O(h)$
Divide & Conquer	Recursive merge	$O(n \log n)$	$O(n)$
QuickHull	Partitioning	Avg: $O(n \log n)$ Worst: $O(n^2)$	$O(n)$

# Applications

- Computer Graphics & Image Processing
- Pattern Recognition & Machine Learning
- Geographic Information Systems (GIS) Mapping

# Convex Hull in Computer Graphics & Image Processing

## Problem

- Objects in images consist of thousands of pixels or vertices
- High computational cost for rendering and collision detection

## Reasoning (Why Convex Hull?)

- Only extreme boundary points define the visible shape
- Interior points do not affect object interaction or visibility

## How It Helps

- Represents objects using a minimal enclosing polygon
- Speeds up collision detection, rendering, and object tracking

## Problem

- Large, multi-dimensional datasets
- Difficulty in identifying class boundaries and anomalies

## Reasoning (Why Convex Hull?)

- All valid data points of a class lie within the region formed by extreme samples
- Points outside the convex hull are statistically unlikely

## How It Helps

- Defines class boundaries geometrically
- Detects outliers and visualizes cluster spread

# Convex Hull in Geographic Information Systems (GIS)

## Problem

- Geographic data (GPS points) are scattered and irregular
- Difficult to determine clear region boundaries

## Reasoning (Why Convex Hull?)

- Convex hull provides the smallest region enclosing all spatial points
- Ensures complete spatial coverage without gaps

## How It Helps

- Boundary mapping of cities, forests, and disaster zones
- Simplifies area and perimeter calculation

# Conclusion

- Convex Hull finds the outer boundary of a set of points.
- All algorithms rely on the orientation test and cross product.
- Different strategies offer different time efficiencies.
- Algorithm choice depends on input size and hull complexity.
- Convex Hull is widely used in GIS, graphics, and collision detection.

# Thank You