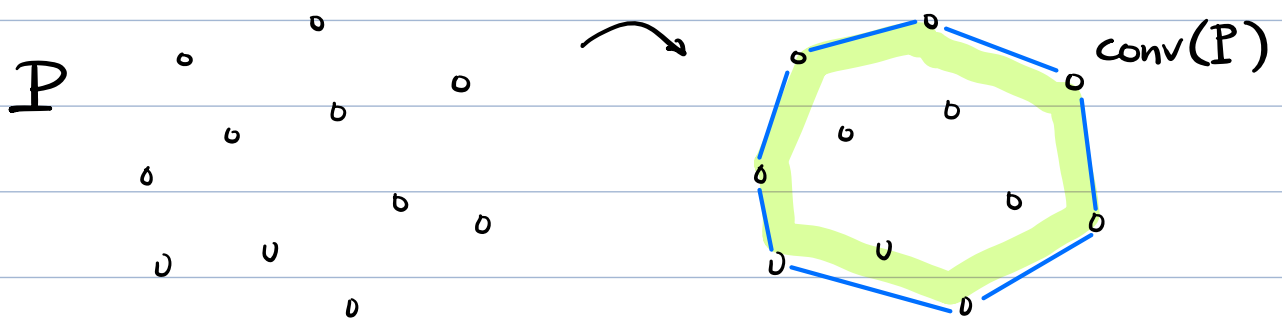


CMSC 754 - Computational Geometry

Lecture 2: Convex Hulls (in the plane)

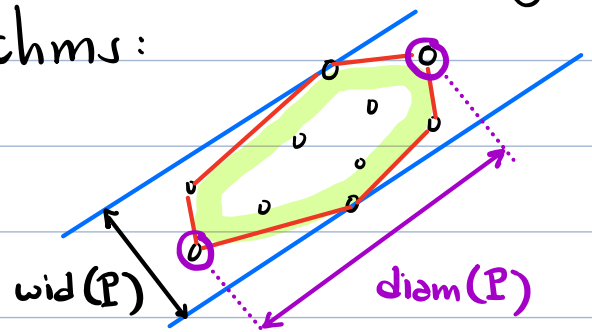
Convex Hull (intuitive definition)

Given a point set P in $\mathbb{R}^2$, imagine snapping a rubber band around the points:



Uses:

- shape approximation (for intersection testing)
- first step in other algorithms:
 - diameter
 - width



Geometric Basics:

Affine Geometry - The geometry of "flats"

Concepts:

- Scalars (Real numbers) $3, -17, \pi, \sqrt{2}, \dots$
- Points (represent positions) $\bullet P \quad \bullet Q$
- Free vectors (direction + magnitude)
(not anchored to the origin) $\vec{u} \quad \vec{v}$

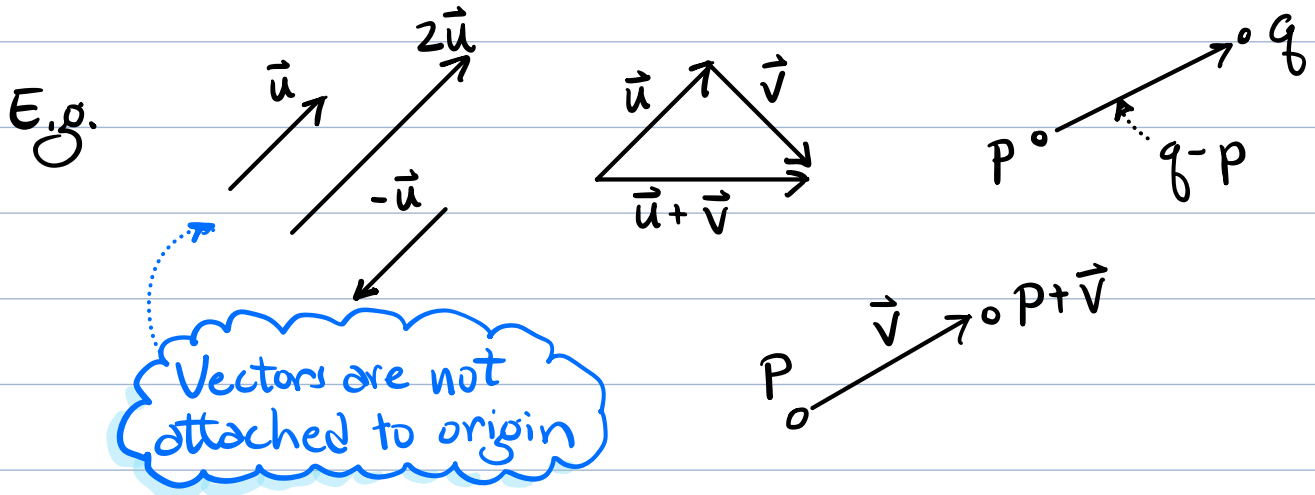
Operations: s = scalar; V = vector; P = point

$s \cdot V \rightarrow V$ - scalar-vector mult

$V + V \rightarrow V$ - vector addition (+ subtraction)

$P - P \rightarrow V$ - point subtraction (result is vect!)

$P + V \rightarrow P$ - point-vector addition



Note: $P + P$ is not allowed!



Can be negative!

Affine Combinations:

Given points p, q and scalars α, β where $\alpha + \beta = 1$

Def:

$\text{aff}(p, q; \alpha, \beta) := p + \beta(q - p)$

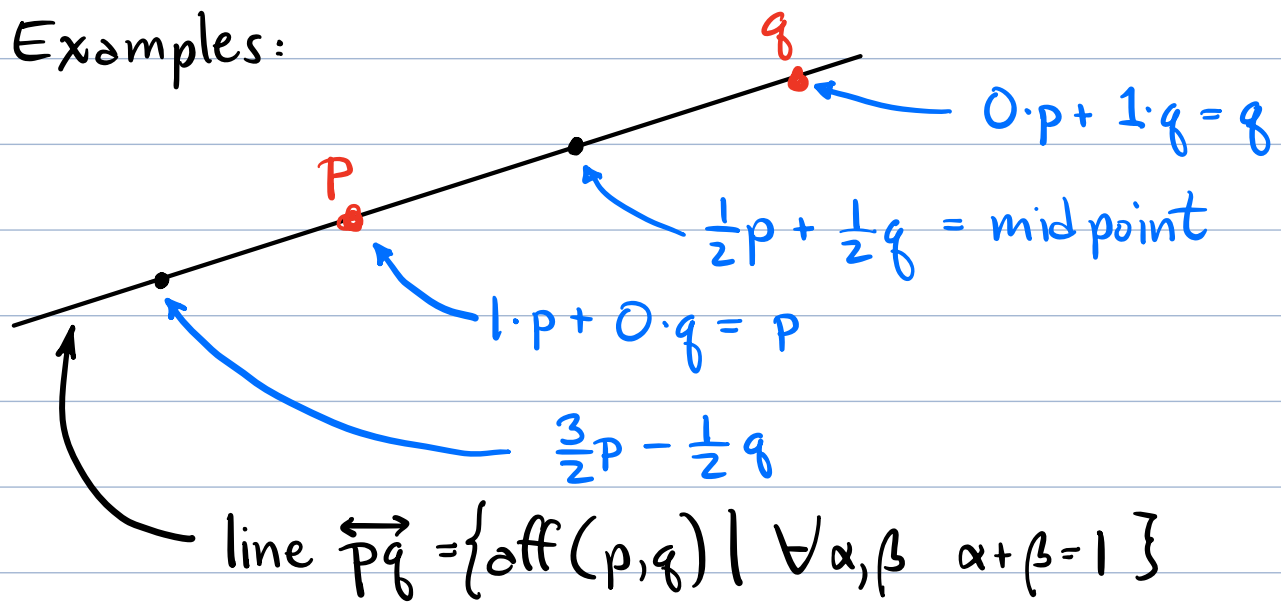
Equivalent to: $p + \beta q - \beta p$

$= (1 - \beta)p + \beta q$

$= \alpha p + \beta q$

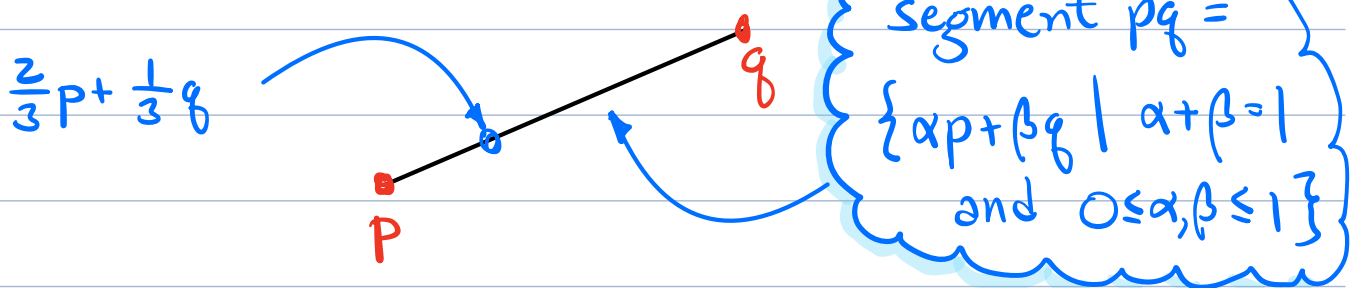
Weighted sum
of p, q

Examples:



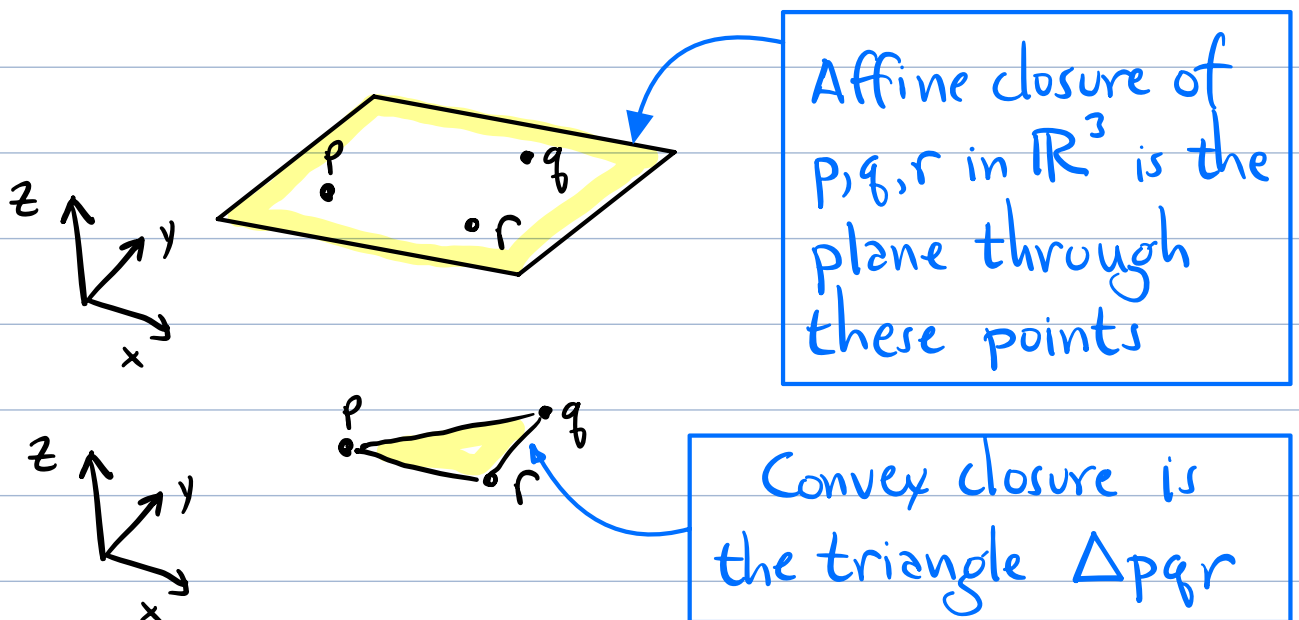
Convex Combination:

Special case when $0 \leq \alpha, \beta \leq 1$



Affine + convex combinations generalize

to multiple points: $\sum \alpha_i p_i$ s.t. $\sum \alpha_i = 1$ (affine)
 $+ 0 \leq \alpha_i \leq 1$ (convex)



... but Affine geometry is limited -
no means for defining lengths or angles

Euclidean Geometry rectifies this by adding...

Inner product - A linear operator that
maps two vectors to a scalar

$$\langle V, V \rangle \rightarrow \mathbb{R}$$

Many varieties, but most common is...

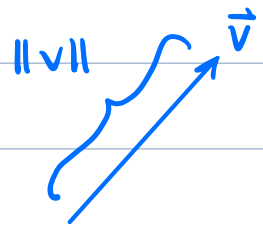
Dot product. Given $\vec{u}, \vec{v} \in \mathbb{R}^d$ $\vec{u} = (u_1, \dots, u_d)$

$$\vec{u} \cdot \vec{v} := u_1 v_1 + \dots + u_d v_d$$

$$= \sum_{i=1}^d u_i v_i$$

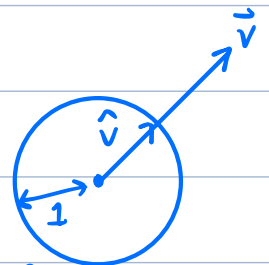
Inner product allows us to define:

Vector length: $\|\vec{v}\| := \sqrt{\vec{v} \cdot \vec{v}}$



Normalization (to unit length):

$$\hat{v} := \vec{v} / \|\vec{v}\| \quad (\text{for } \vec{v} \neq 0)$$



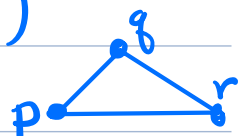
Distance: Given points p, q

$$\text{dist}(p, q) := \|q - p\|$$



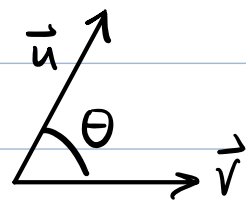
- Symmetric: $\text{dist}(p, q) = \text{dist}(q, p)$

- Triangle inequality: $\text{dist}(p, r) \leq \text{dist}(p, q) + \text{dist}(q, r)$



Angles: $\vec{u}, \vec{v} \neq 0$

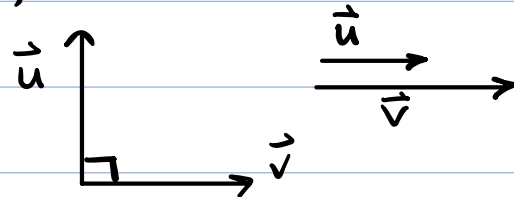
$$\text{ang}(\vec{u}, \vec{v}) := \cos^{-1}(\hat{u} \cdot \hat{v})$$



Perpendicularity / Parallelity: $\vec{u}, \vec{v} \neq 0$

$$u \perp v \Leftrightarrow \vec{u} \cdot \vec{v} = 0$$

$$u \parallel v \Leftrightarrow \|\hat{u} \cdot \hat{v}\| = 1$$



Flats (lines + hyperplanes)

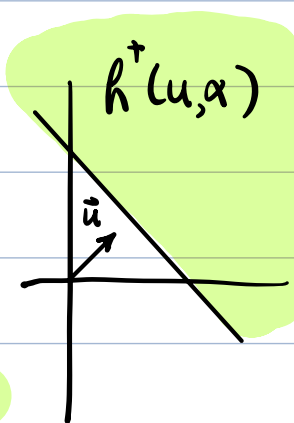
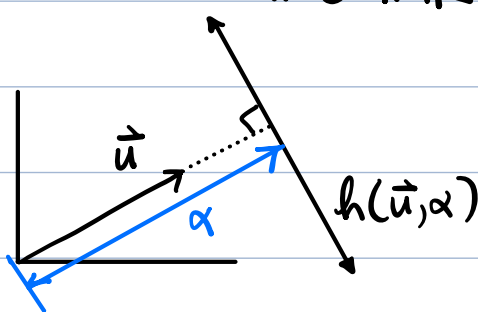
Given vector $\vec{u} \neq 0$ + scalar α , define flat

$$h(\vec{u}, \alpha) := \{p \mid p \cdot \vec{u} = \alpha\}$$

- line in $\mathbb{R}^2$, hyperplane in $\mathbb{R}^d$

Picture

(when $\|\vec{u}\|=1$)

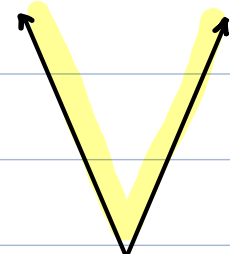
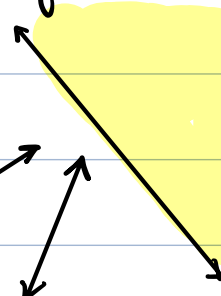
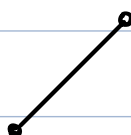
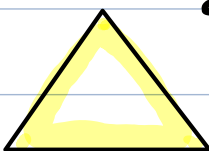
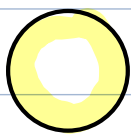
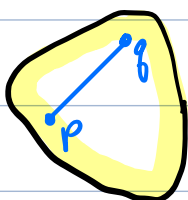


Halfspace: The region on one side of

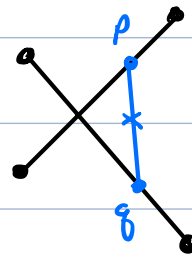
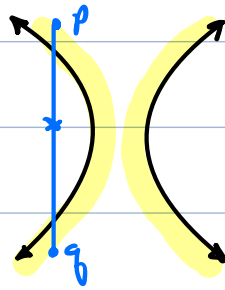
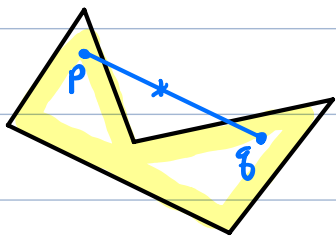
a hyperplane: $h^+(\vec{u}, \alpha) := \{p \mid p \cdot \vec{u} \geq \alpha\}$

Convex: A point set $K \subseteq \mathbb{R}^d$ is convex if $\forall p, q \in K$, all convex combinations of $p+q$ lie within K .

convex {



not convex



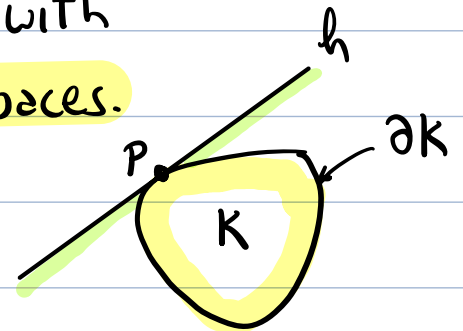
Intersection of convex sets is always convex

Union of convex sets need not be convex

Supporting Hyperplane:

Given a convex K and any point $p \in \partial K$.
 $\exists$ hyperplane h passing through p with K lying in one of h 's closed halfspaces.

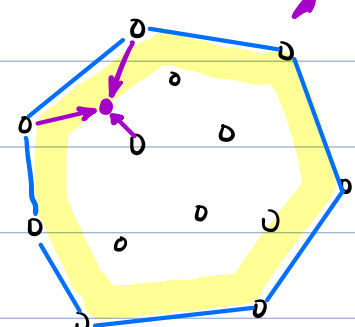
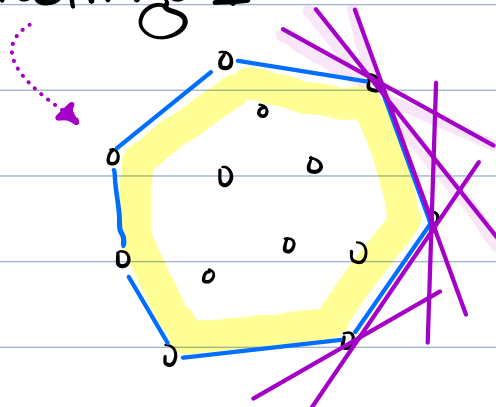
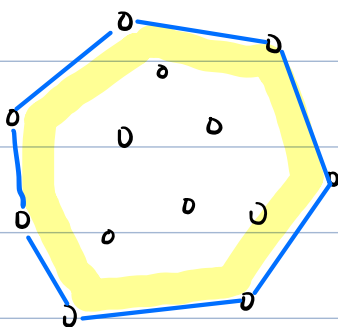
Means:
Boundary of K



Convex Hull:

Given a point set P in $\mathbb{R}^d$, the convex hull, $\text{conv}(P)$, is the smallest convex set containing P .

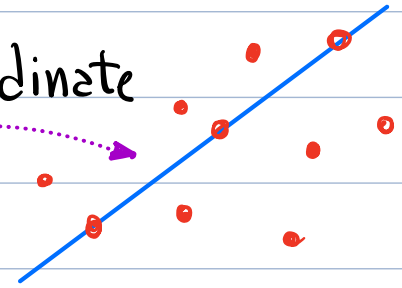
Equiv: - Set of all convex combinations of P
 - Intersection of all halfspaces containing P



General Position:

Geometric algorithms are often complicated by (rare) geometric degeneracies

- E.g.
- points having the same coordinate
 - ≥ 3 collinear points
 - ≥ 4 cocircular points



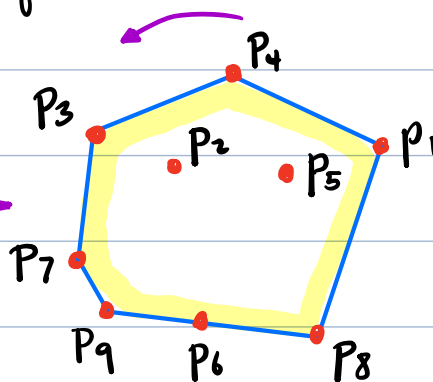
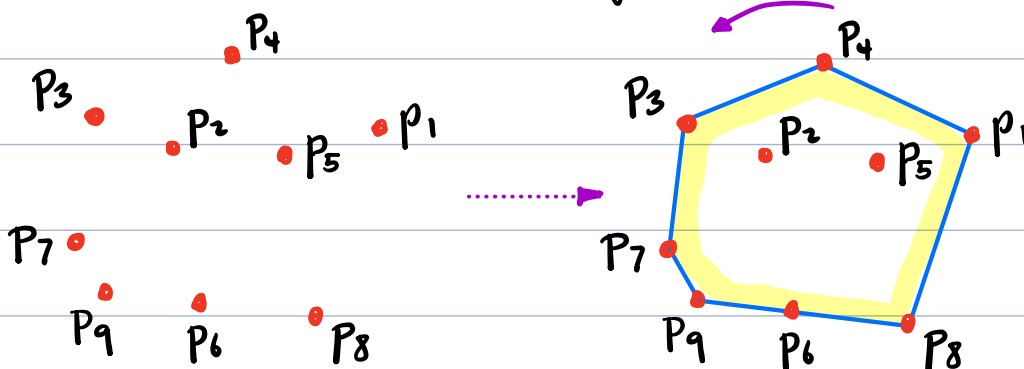
To simplify our algorithms, we assume inputs are perturbed (infinitesimally) to avoid these.

We say such inputs are in "general position"

(Planar) Convex Hull problem - Given n points

$P = \{p_1, \dots, p_n\} \subseteq \mathbb{R}^2$, where $p_i = (x_i, y_i)$, compute convex hull of P , $\text{conv}(P)$.

Output: Cyclic sequence of vertices on the hull



Output:

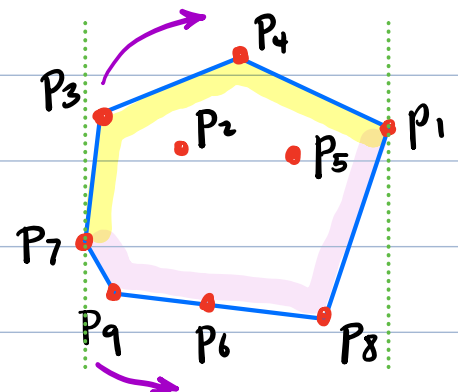
$\langle 4, 3, 7, 9, 8, 1 \rangle$

Note: 6 not included. Not a vertex

Alternative output:

Upper hull + Lower hull

$\langle 7, 3, 4, 1 \rangle + \langle 7, 9, 6, 8, 1 \rangle$



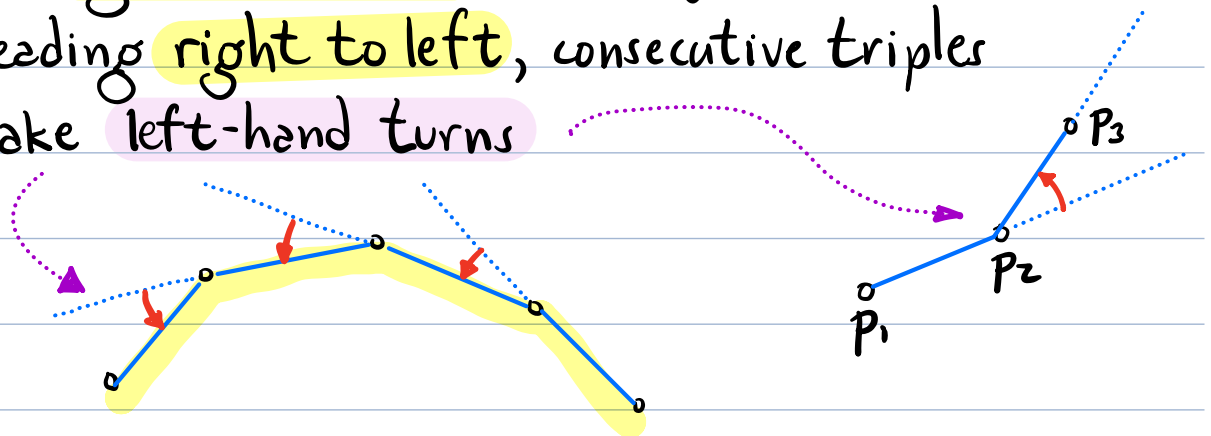
Graham's Scan - An $O(n \log n)$ algorithm

- Computes upper + lower hulls separately
- Upper hull
 - Sort points by x-coords.
 - Add each to upper hull
 - Remove points no longer on hull
- Lower hull (symmetrical)

How?

Observations:

- The rightmost point is always on the hull
- Reading right to left, consecutive triples make left-hand turns



Incremental approach:

Store vertices of current upper hull on stack S

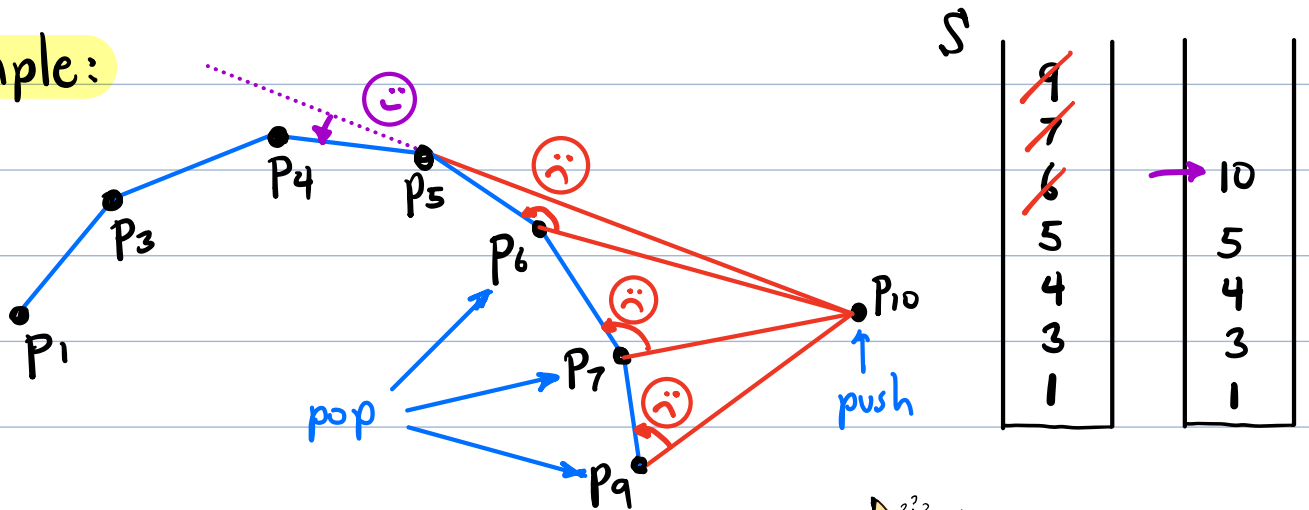
For each new point p_i (left to right)

While $(p_i, S[top], S[top-1])$ do not

make left turn - pop S

Push p_i

Example:



But how to test for left turn?

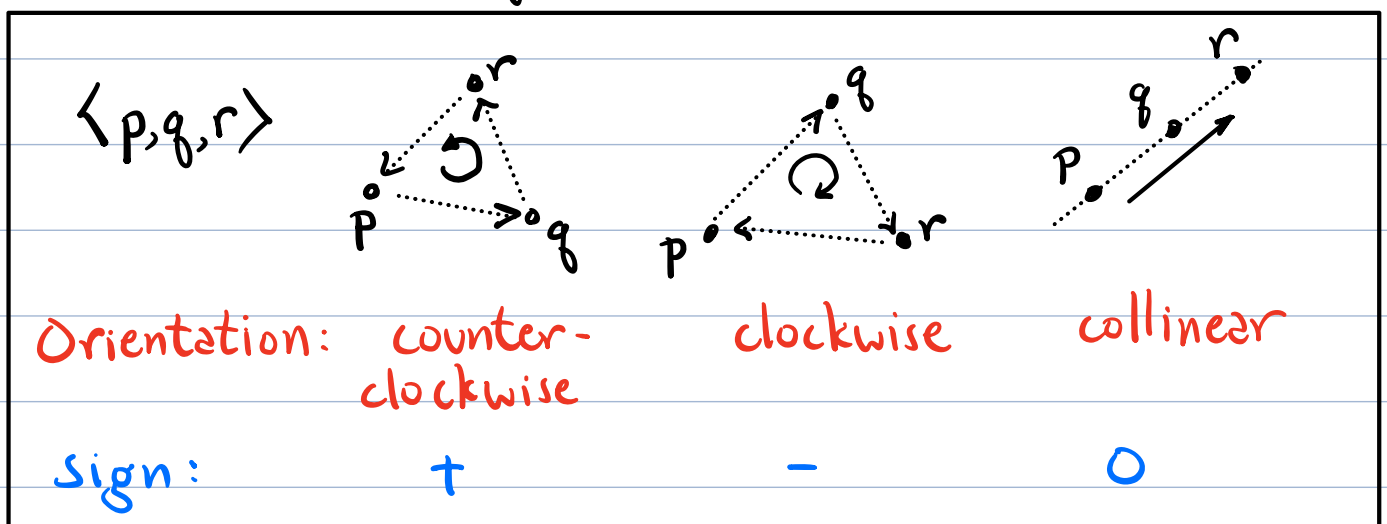


Orientation Test: For $p, q \in \mathbb{R}^1$, we have either:

	Order:	Sign:	Orient:
	$p < q$	$q - p > 0$	+
	$p = q$	$q - p = 0$	0
	$p > q$	$q - p < 0$	-

How to define "order" in $\mathbb{R}^2$?

Ans: Order (orientation) is a property of a sequence of 3 points:



How to compute?

$\text{orient}(p, q, r) = \text{sign det}$

$$\begin{bmatrix} 1 & p_x & p_y \\ 1 & q_x & q_y \\ 1 & r_x & r_y \end{bmatrix}$$

Note: (Cyclic) order of points is critical!

Higher dimensions - Orientation in $\mathbb{R}^d$ is
a property of a sequence of $d+1$ points

Graham's Scan (upper hull only)

Sort points by increasing x-coords $\rightarrow \langle p_1, \dots, p_n \rangle$

Push p_1, p_2 onto S

for $i \leftarrow 3$ to n

while ($\text{orient}(p_i, S[\text{top}], S[\text{top}-1]) \leq 0$)

pop(S)

push p_i

Right turning
or straight

Correctness:

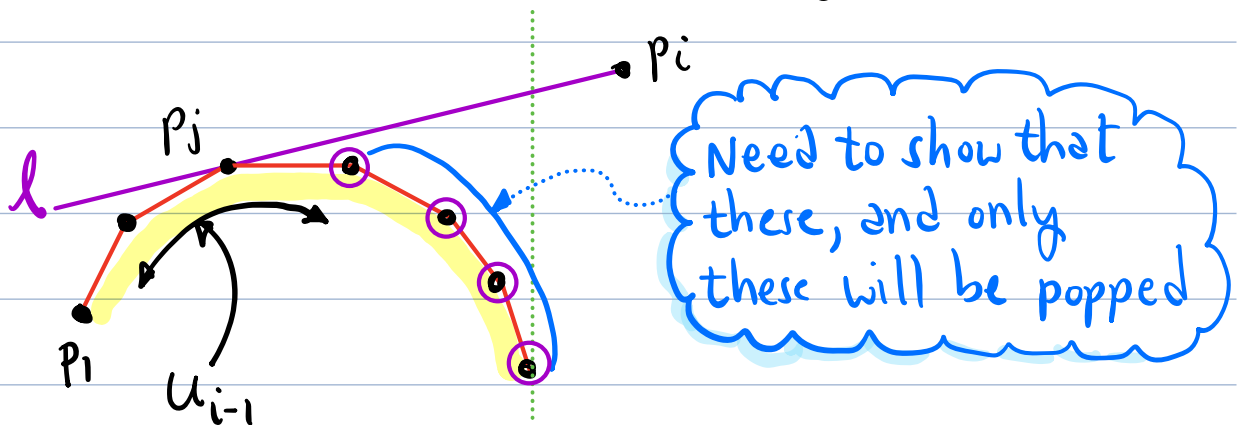
Lemma: After processing p_i , S contains upper hull of $\{p_1, \dots, p_i\}$

Proof: By induction on i

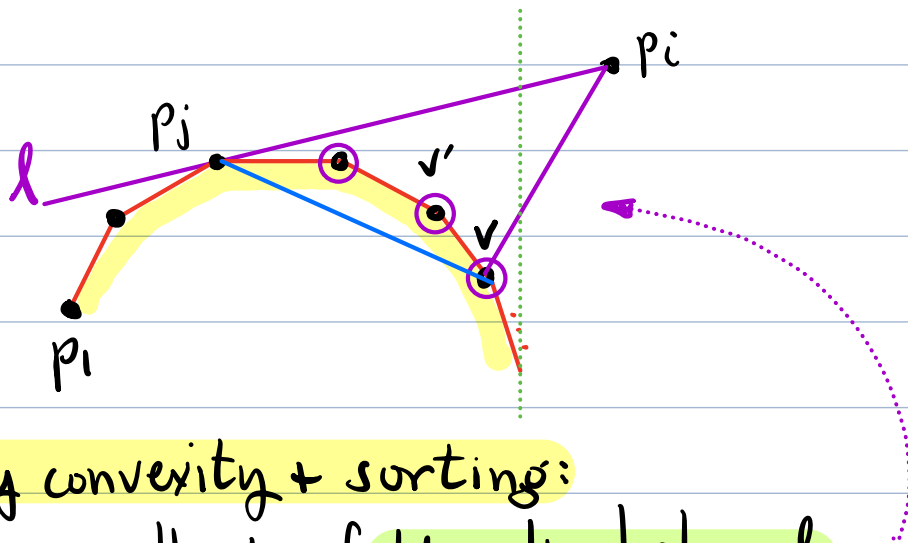
Basis: $i = 1, 2 \rightarrow$ Trivial

Step: For $i \geq 3$, let U_i denote vertices of upper hull of $\{p_1, \dots, p_i\}$

- By induction U_{i-1} is correct up to p_{i-1}
- Let p_j be vertex prior to p_i on U_i
 $\Rightarrow$ line $l := \overleftrightarrow{p_j p_i}$ is a supporting line for U_i



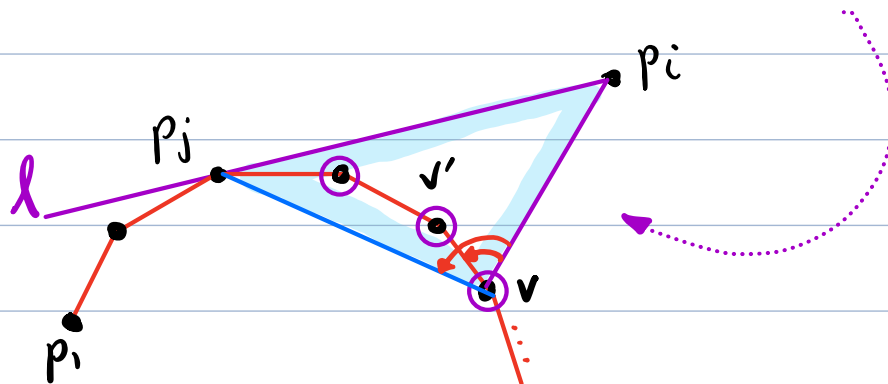
Let $v := S[\text{top}] + v' := S[\text{top}-1]$



By convexity + sorting:

- all pts of U_{i-1} lie below l
- all pts of U_{i-1} from p_j to v are above l

$\Rightarrow$ All points of S from p_j to v lie in $\Delta p_j v p_i$


$$\Rightarrow \angle p_i v v' \leq \angle p_i v p_j \leq \pi$$
$$\Rightarrow \text{orient}(p_i, v, v') = \text{orient}(p_i, S[\text{top}], S[\text{top}-1]) \leq 0$$

$\Rightarrow v$ is popped off stack

The popping repeats until $v = p_j$

$\Rightarrow$ Final stack $S = \langle p_1, \dots, p_j, p_i \rangle = U_i$



Running Time:

- Sort by x-coords - $O(n \log n)$

- Popping time?

- Outer loop $i \leftarrow 3$ to n

- Inner loop

- Could pop up to $i-2$ points

- Total:

$$\leq \sum_{i=3}^n i-2 = O(n^2)$$



Why is this wrong?

- Each point can be popped only once.
- Let d_i = no. of points popped when p_i is processed
- $\sum d_i \leq n$
- Total inner loop time:

$$\sum_{i=3}^n 1 + d_i \leq \sum_{i=3}^n d_i + (n-2) \leq O(n)$$



Total time: $O(n \log n) + O(n) = O(n \log n)$

Summary:

- Affine + Euclidean Geometry
- Affine + Convex combinations
- Geometric operations
 - inner product
 - orientation test
- Convexity + convex hulls
- Graham's Algorithm