

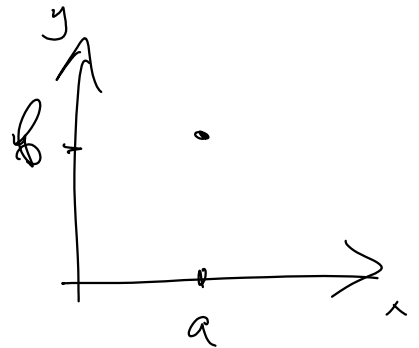
Computational Geometry

Algorithms for problems that involve geometric objects:

- points
- lines
- line segments
- planes (in 3-D & above)

Geometry in 2-D

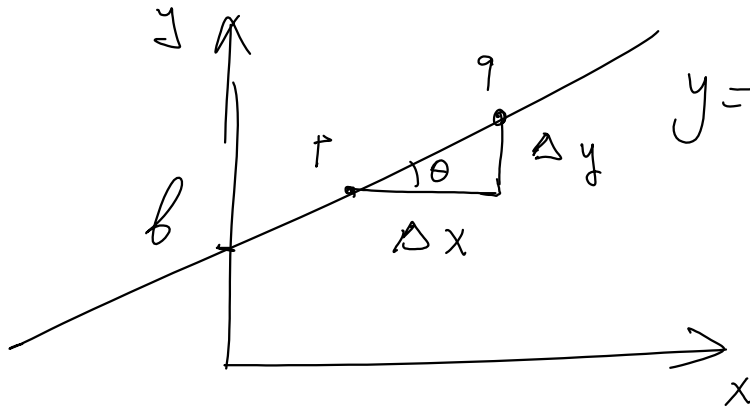
Points : $P = (a, b)$



lines : $ax + by + c = 0$

$$y = \overset{\circ}{m}x + \overset{\circ}{b} \leftarrow \text{non-vertical lines}$$

\uparrow \uparrow
 slope y-intercept



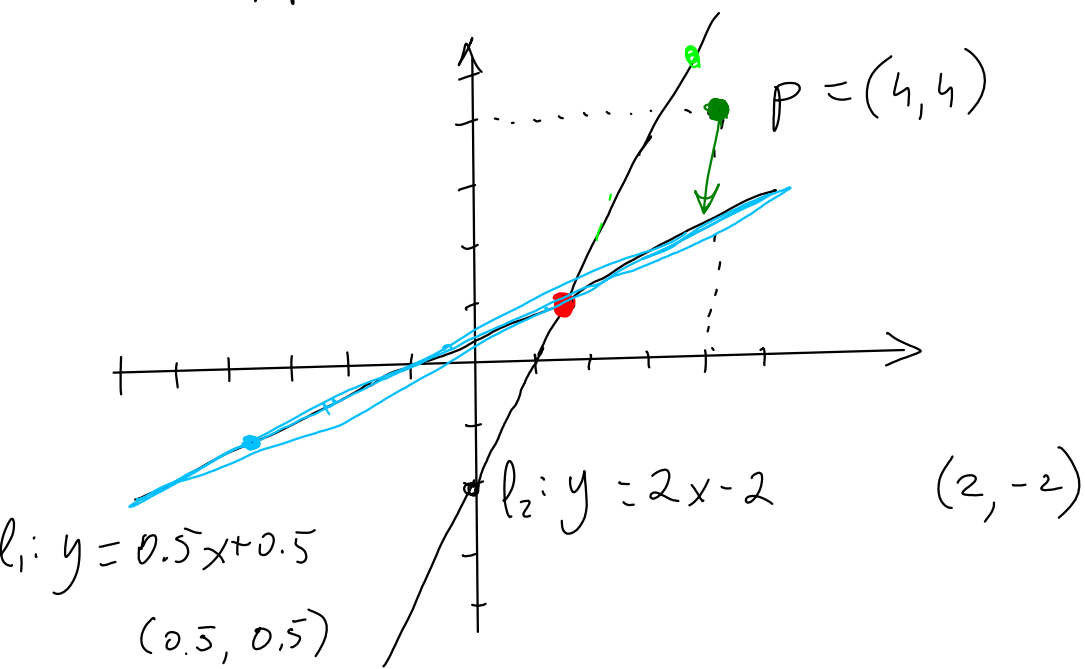
$$m = \frac{\Delta y}{\Delta x} = \tan \theta$$

$$l = (m, b)$$

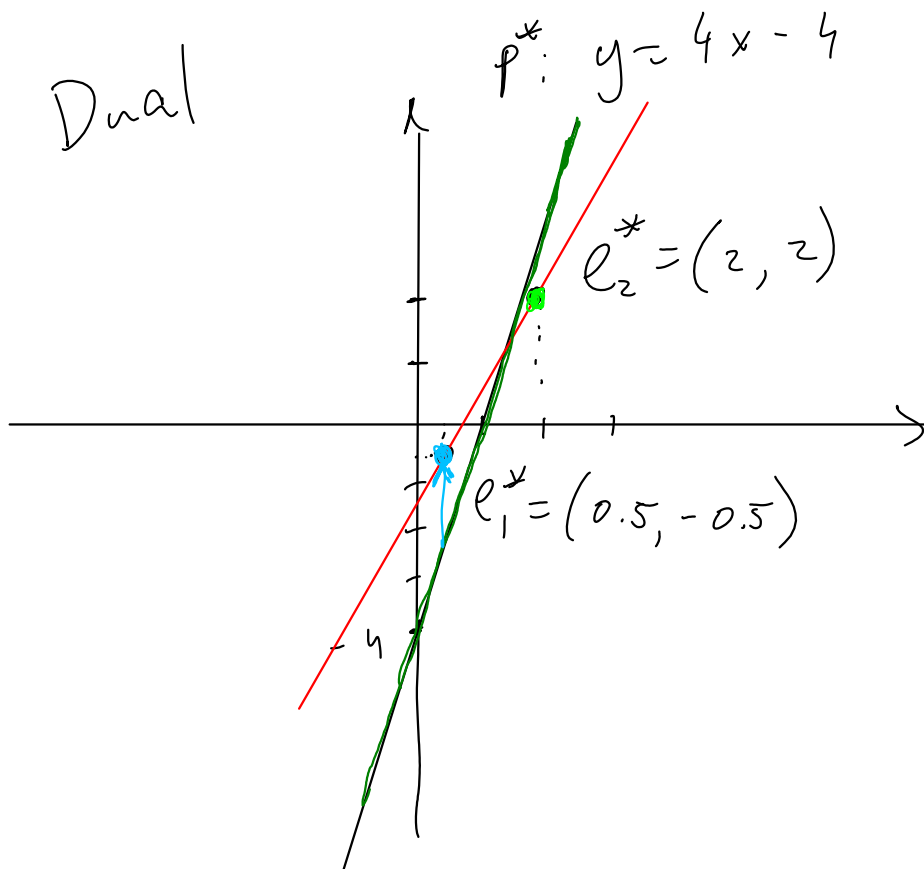
Primal Space		Dual Space
$p = (a, b)$	\implies	$l^* = (a, -b)$ $y = ax - b$

$l = (a, b)$	\implies	$p^* = (a, -b)$
$y = ax + b$		

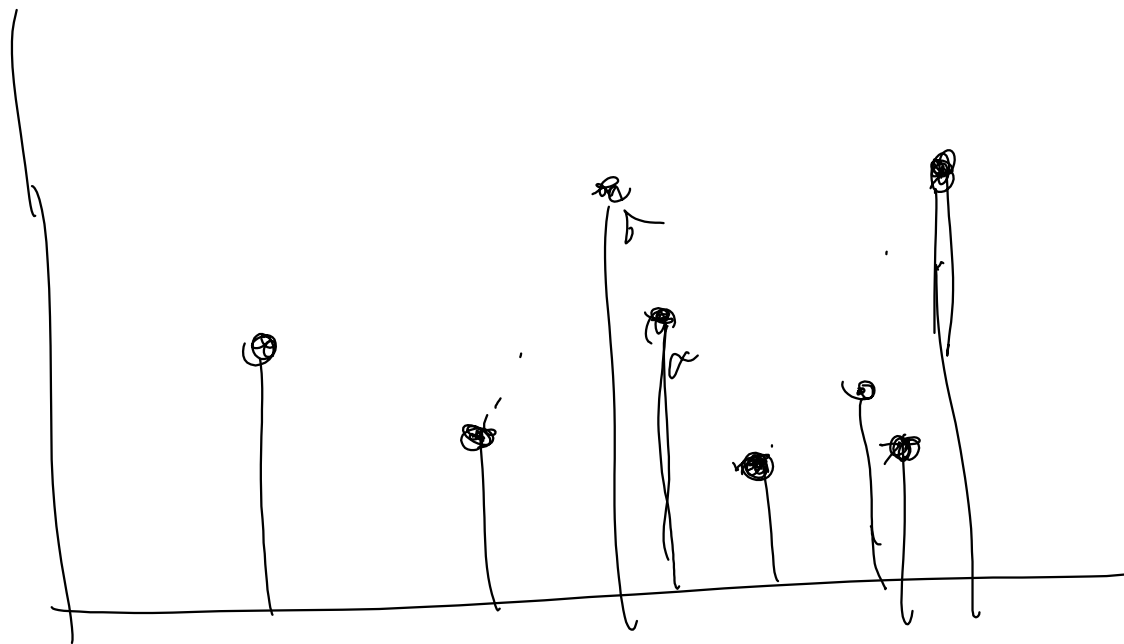
Primal



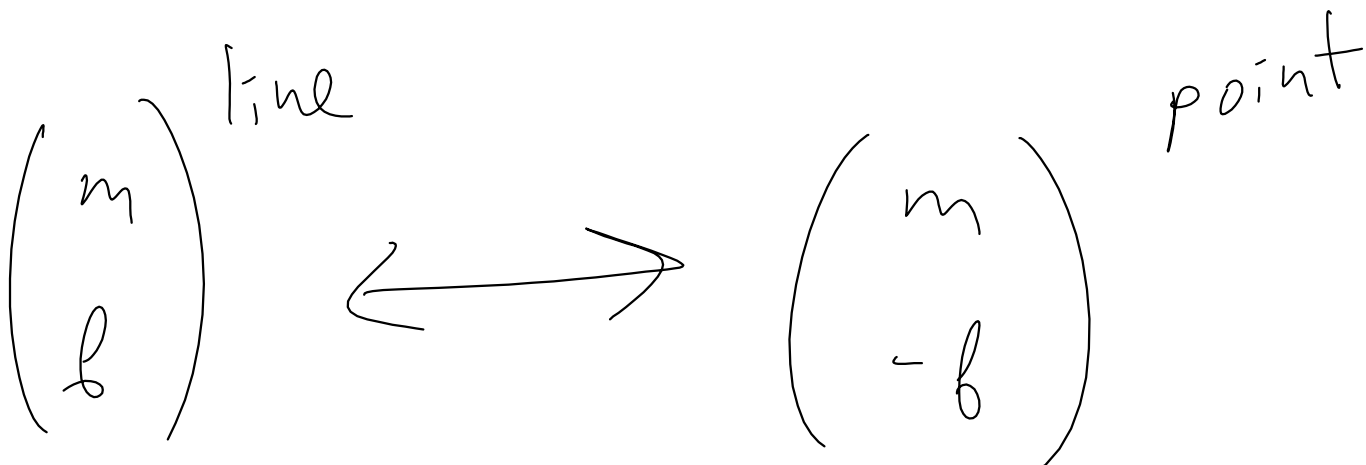
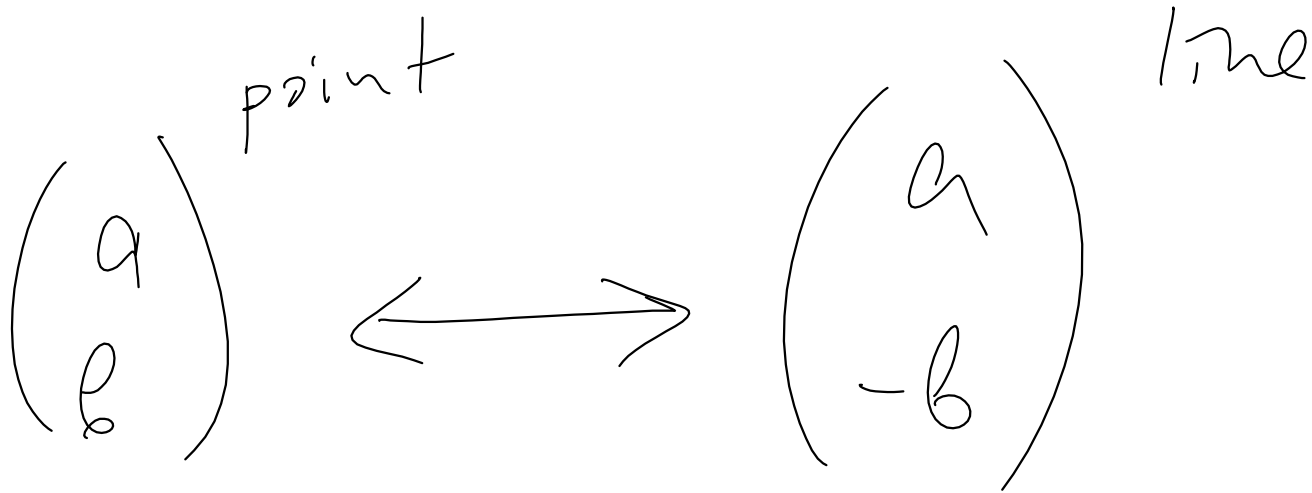
Dual



Given a point p & a line l in
primal space, p lies above l
iff p^* lies above e^* in dual space

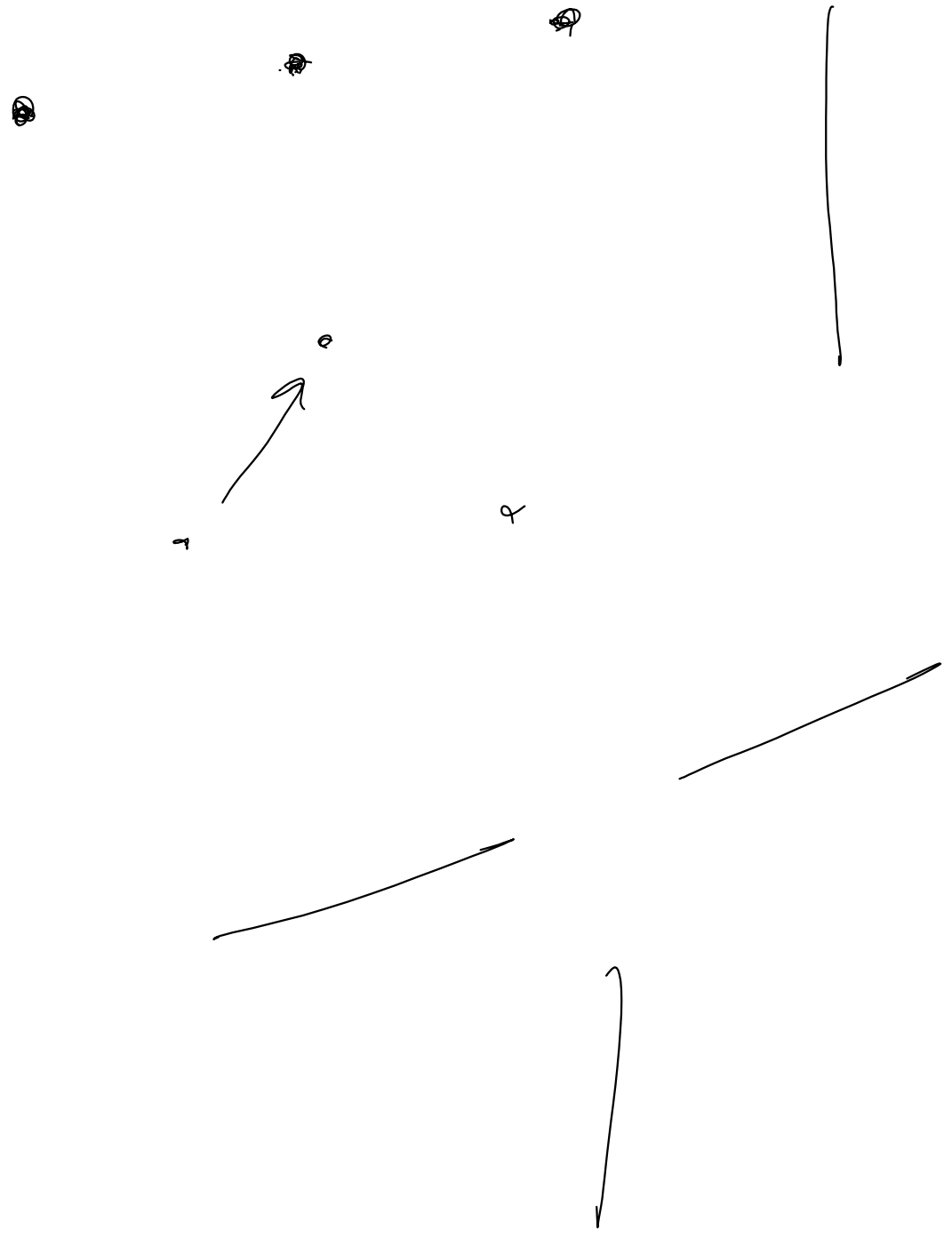


E.g. visibility counting problem



General Position

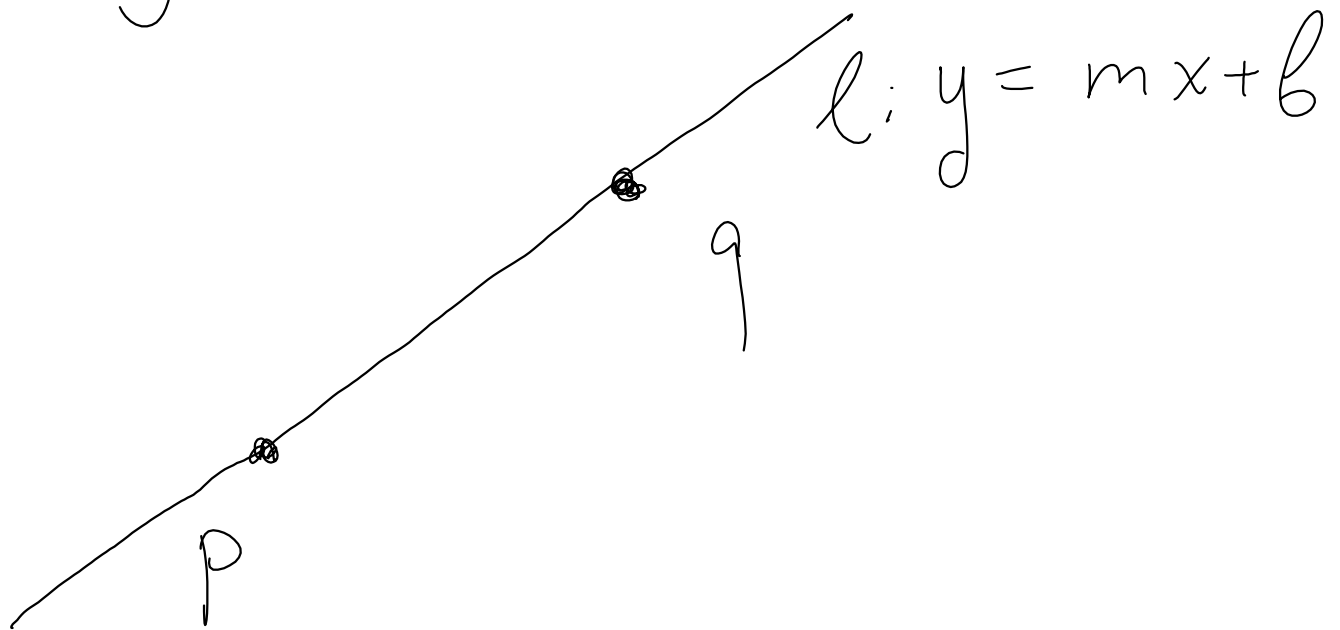
- No three points are co-linear



Given two points $p = (\underline{p}_x, \underline{p}_y)$

$$q = (\underline{q}_x, \underline{q}_y).$$

They define a line



$$\left\{ \begin{array}{l} p_y = m p_x + b \\ q_y = m q_x + b \end{array} \right. \Rightarrow \begin{array}{l} \underline{\underline{b = p_y - m p_x}} \\ \underline{\underline{b = q_y - m q_x}} \end{array}$$

$$P_y - m P_x = Q_y - m Q_x$$

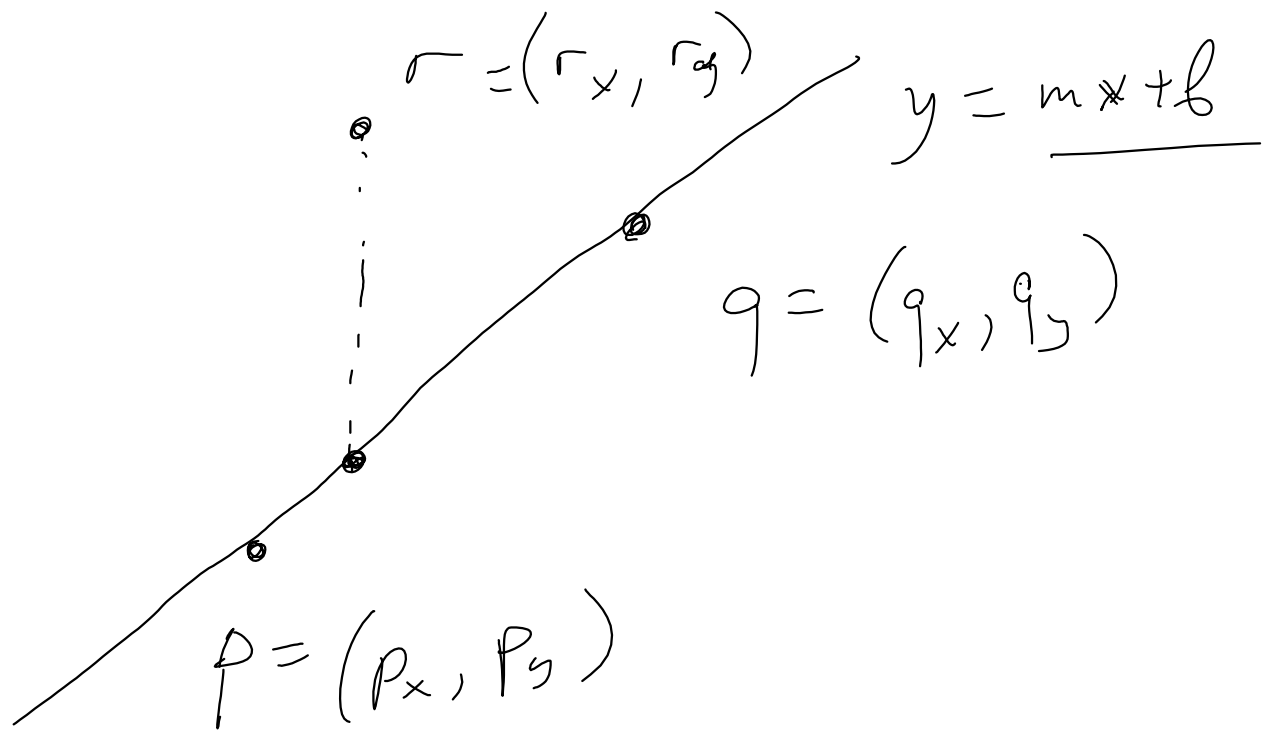
$$m (P_x - Q_x) = P_y - Q_y$$

$$m = \frac{P_y - Q_y}{P_x - Q_x}$$

$$b = P_y - \frac{P_y - Q_y}{P_x - Q_x} \cdot P_x$$

$$= \frac{\cancel{P_x P_y} - P_y Q_x - \cancel{P_x P_y} + P_x Q_y}{P_x - Q_x}$$

$$= \frac{P_x Q_y - P_y Q_x}{P_x - Q_x}$$



$$r_y \geq m \cdot r_x + b$$

$$\geq \frac{P_y - q_y}{P_x - q_x} \cdot r_x + \frac{P_x q_y - P_y q_x}{P_x - q_x}$$

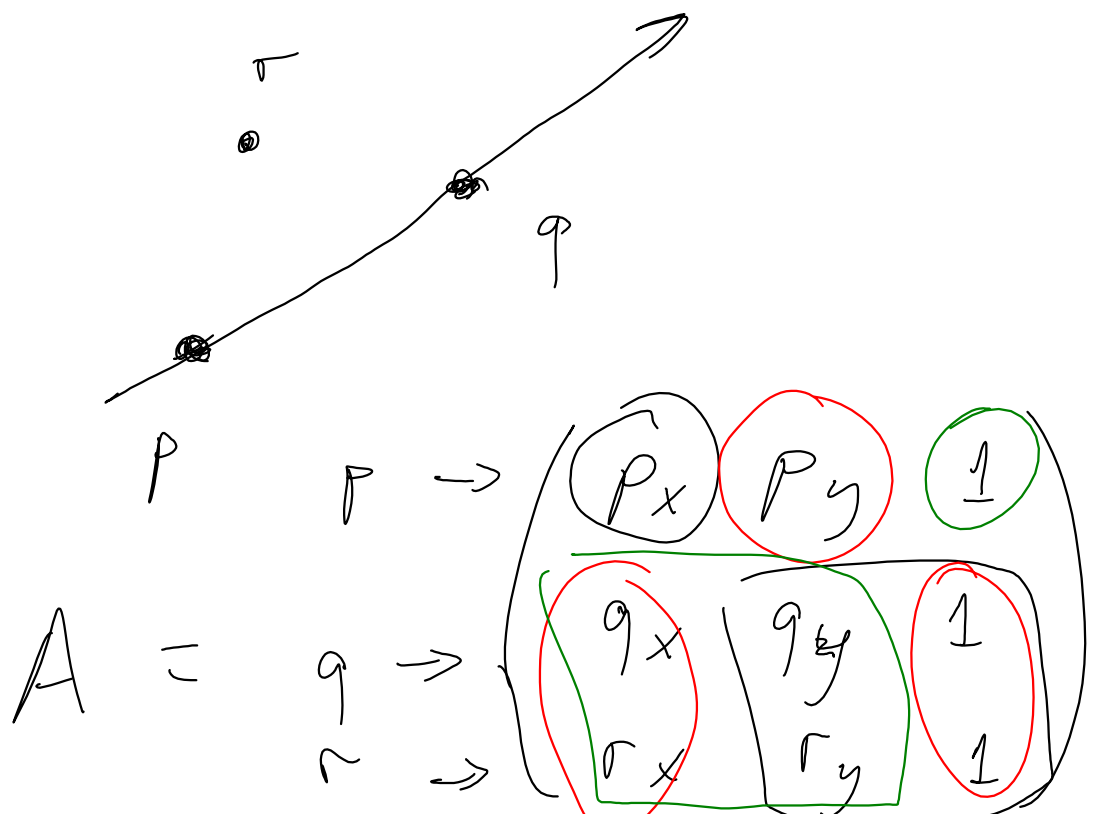
Robustness (rounding errors)

$$\text{if } x = y \Rightarrow \text{if } |x - y| \leq \varepsilon$$

.....

$$r_y (p_x - q_x) \stackrel{>}{=} (p_y - q_y) \cdot r_x + p_x q_y - p_y q_x$$

$\Leftrightarrow r$ is above \overline{pq}
 or below



r is to the left or right of \vec{pq}

iff $|A| \begin{cases} < \\ = \\ > \end{cases} 0$

$$|A| = P_x \cdot \begin{vmatrix} q_y & 1 \\ r_y & 1 \end{vmatrix} - P_y \cdot \begin{vmatrix} q_x & 1 \\ r_x & 1 \end{vmatrix} + 1 \cdot \begin{vmatrix} q_x & q_y \\ r_x & r_y \end{vmatrix}$$

$$P_x (q_y - r_y) - P_y (q_x - r_x) + q_x r_y - q_y r_x$$

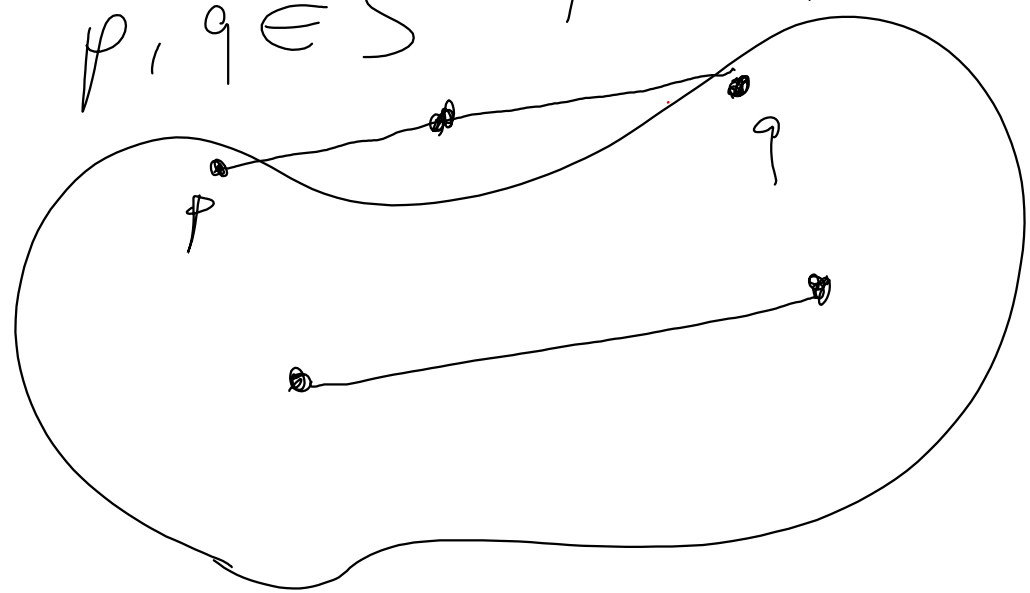
$$|A| = \begin{vmatrix} (P_x - r_x) & (P_y - r_y) \\ (q_x - r_x) & (q_y - r_y) \end{vmatrix}$$

$$= (P_x - r_x)(q_y - r_y) - (P_y - r_y)(q_x - r_x)$$

\uparrow \uparrow

Def'n: An object S is convex
 iff for every pair of
 points $p, q \in S$ the segment

$$\overline{p, q} \in S$$



$$\alpha p + (1 - \alpha) q \in S \text{ for all } 0 \leq \alpha \leq 1$$

Given points \vec{p} & \vec{q}

the line segment \overline{pq}
is a set of all points

defined by $\alpha \vec{p} + (1-\alpha) \vec{q}$

for all $0 \leq \alpha \leq 1$

$$P = \begin{pmatrix} p_x \\ p_y \end{pmatrix} \quad Q = \begin{pmatrix} q_x \\ q_y \end{pmatrix}$$

$$\alpha \vec{p} + (1-\alpha) \vec{q} =$$

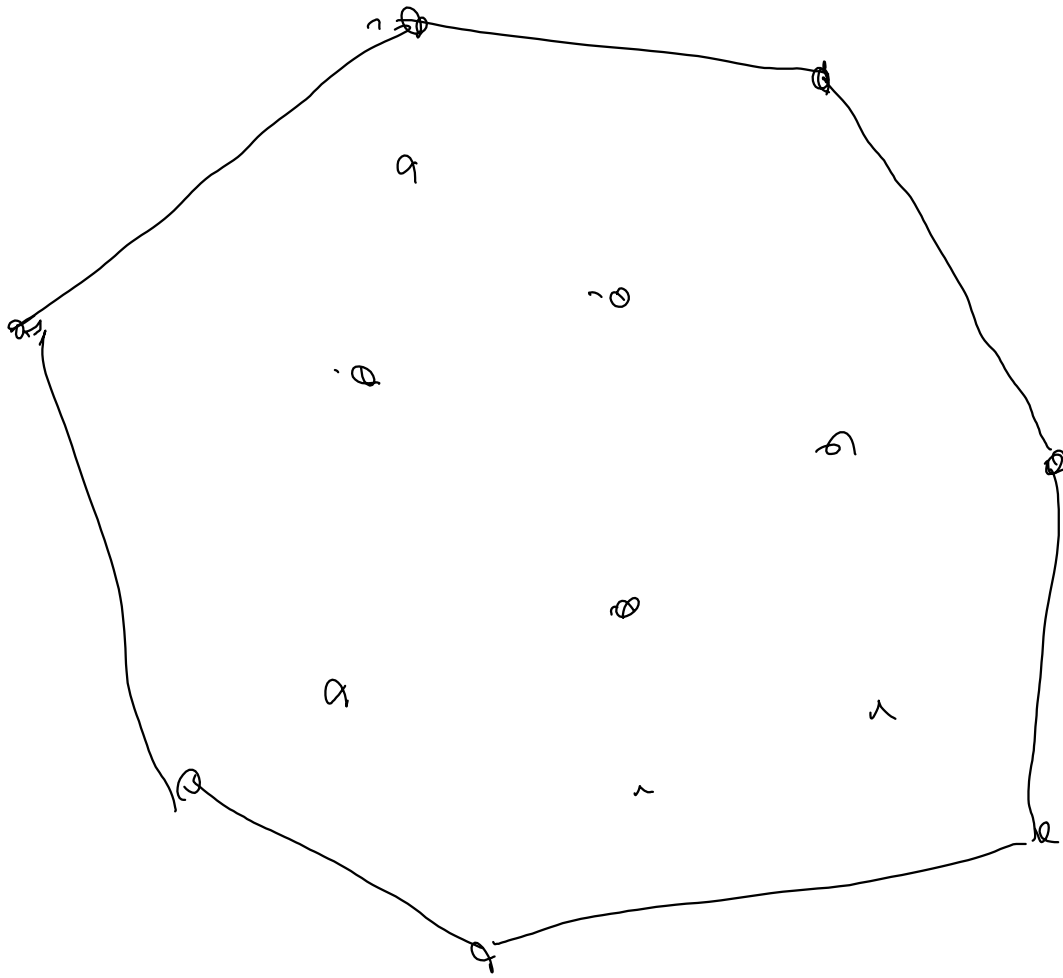
$$\begin{pmatrix} \alpha p_x + (1-\alpha) q_x \\ \alpha p_y + (1-\alpha) q_y \end{pmatrix}$$

Def'n: (Convex Hull)

Given a set of points P in \mathbb{R}^2 a convex hull

{ is a minimal convex set that contains P

{ is an intersection of all convex sets that contain P .

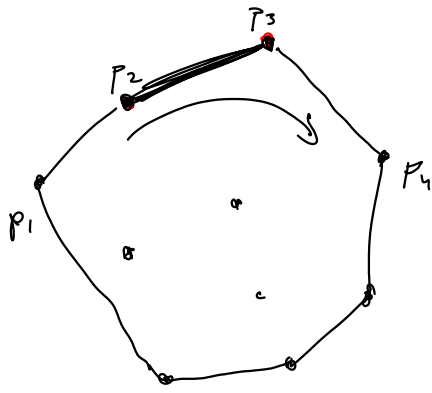


Computing convex hull:

Input: n points in \mathbb{R}^2

Output: set of points on the boundary of convex hull in clockwise order

Convex Hull : ^{Maximal} Intersection of all half planes that contains all points

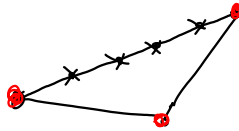


Convex Hull:

Input: set of points

Output: set of points on the convex hull in clockwise direction

Trivial alg:



$O(n^3)$ { For every pair of points p_i, p_j : $\leftarrow O(n^2)$
 if half planes H_{ij} or H_{ji} contain all $O(n)$ points, then include $\overline{p_i, p_j}$ into Convex Hull

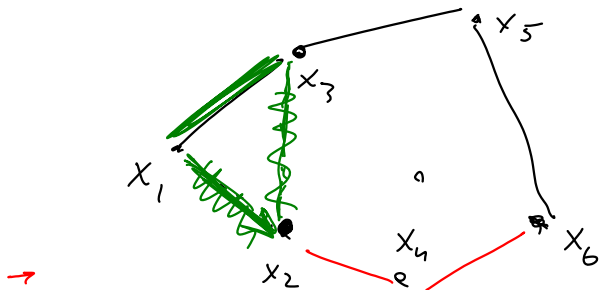
$O(n^2)$ { sort points on the Convex Hull in clockwise order



$$(n-2) + (n-3) + (n-4) \dots 1 = O(n^2)$$

Convex Hull in $O(n \log n)$

using plane sweep technique



sort points by x-coord,
 process these sorted points in this order

Upper Convex Hull:

→ Sort points by x-coordinate (P_1, P_2, \dots, P_n)

(inlogn) Add $\overrightarrow{P_1, P_2}$ to "convex hull"

for $i = 3$ to n

$P = P_i$

~~if~~ while P lies to the left of $\overrightarrow{P_{i-2}, P_{i-1}}$ then

remove P_{i-1} & add P_i

$S.push(P_1); S.push(P_2)$

for $i = 3$ to n \neq

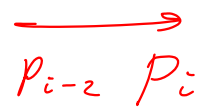
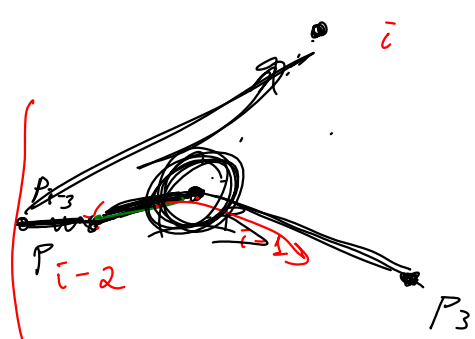
$P = P_i$

$P' = S.pop()$

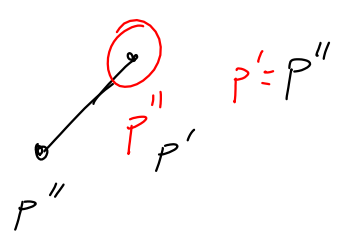
$P'' = S.pop()$

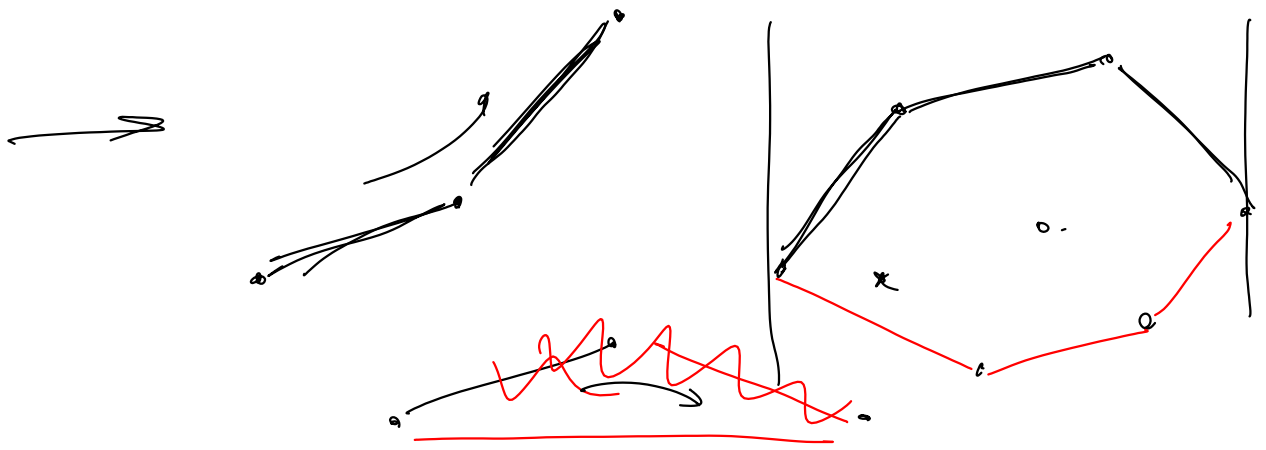
{ while P lies to the left of $\overrightarrow{P''P'}$
 $P' = P''$
 $P'' = S.pop()$ \neq

$S.push(P'')$; $S.push(P')$; $S.push(P)$



$O(n)$





Loop Invariant Prior to i th iteration,
 Stack S contains $n+1$ all points of the
 upper convex hull of points $p_1 \dots p_{i-1}$
 $p_{n+1-i} = p_n$

Termination: stack S will contain all points of
 upper convex hull of points $p_1 \dots p_n$

Maintenance:

Lower bound
proof

Convex Hull takes $\Omega(n \log n)$ time

CH-sort ($A[1 \dots n]$)

for $i = 1$ to n

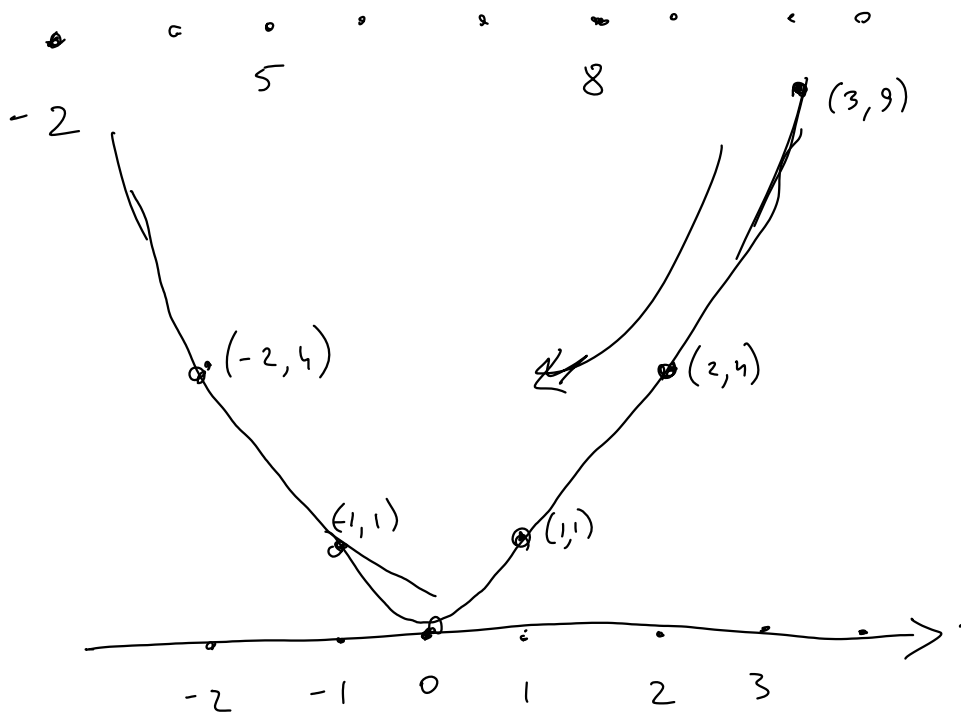
$P[i] = \text{new Point}(A[i], A[i] \cdot A[i])$

$c = \text{lower CH}(P[1 \dots n])$

for $i = n$ down to 1
print $C[i].x$

$(-2, (-2)^2)$

$(5, 5^2)$

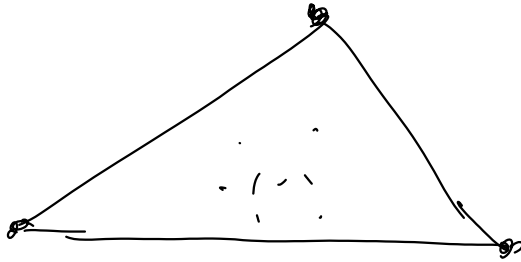


This proof is an example of reduction
of sorting to convex hull

Best Convex Hull algorithm runs in

$O(n \log h)$ time,

where h is # of points on
convex hull.



output-sensitive
algorithms