



Advanced Algorithms

Winter term 2019/20

Lecture 3. 2D Linear Programming via sweep-lines and randomization

Source: **CG: A&A** §**4**

Maximizing Profit

You are the boss of a small company that produces two products, P_1 and P_2 . If you produce x_1 units of P_1 and x_2 units of P_2 , your profit in \in is

$$G(x_1, x_2) = 300x_1 + 500x_2$$

Your production runs on three machines M_A , M_B , and M_C with the following capacities:

 M_A : $4x_1 + 11x_2 \le 880$ M_B : $x_1 + x_2 \le 150$ M_C : $x_2 < 60$

Which choice of (x_1, x_2) maximizes your profit?

The Answer

 χ_2

150

30,000 E





$$M_B: x_1 + x_2 \le 150$$

$$M_C: x_2 \leq 60$$

$$x_1 \geq 0$$

$$x_2 \ge 0$$

$$Ax \leq b$$

$$x \ge 0$$

linear objective fct.: maximize $c^{T}x$

$$G(x_1, x_2) = 300x_1 + 500x_2$$

= $(300, 500) \binom{x_1}{x_2}$



$$G(110,40) = 53,000$$

- = maximum value of objective fct. given constraints
- $= \max\{c^{\mathrm{T}}x \mid Ax \leq b, x \geq 0\}$

50 100 150 200
$$x_1$$
 "iso-profit line" (orthogonal to $\binom{300}{500}$)

Definition and Known Algorithms

Given a set H of n halfspaces in \mathbb{R}^d and a direction c, find a point $x \in \bigcap H$ such that cx is maximum (or minimum).

Many algorithms known, e.g.:

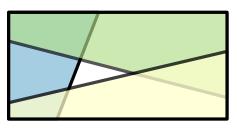
- Simplex [Dantzig '47]
- Ellipsoid method [Khatchiyan '79]
- Inner-point method [Karmakar' 84]

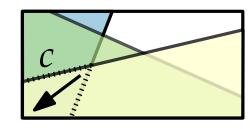
Good for instances where *n* and *d* are large.

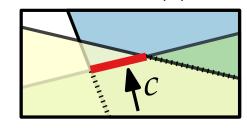
We consider d = 2.

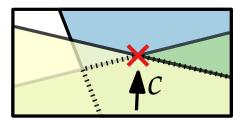
VERY important problem, e.g., in Operations Research.

["Book" application: casting]









 $\bigcap H = \emptyset$ $\bigcap H$ unbind. in dir. c

set of optima: segment vs. point

H bounded.

First Approach

- compute $\bigcap H$ iteratively
- walk $\partial (\cap H)$, find vertex $x \le w / cx$ maximum, O(n) time

```
IntersectHalfplanes(H)

Let H = (h_1, ..., h_n)

C \leftarrow h_1

foreach i from 2 to n do

C \leftarrow C \cap h_i

return C
How??
```

C := chain of line $segments (s_1, ..., s_t)$ Walk around C to find $s_j, s_{j'} \in C$ intersecting h_i Update C

Running time: $T_{IH}(n) = n \cdot O(n)$

Total Time: $O(n^2)$:(

Exercise: Compute $C \cap h_i$ faster.

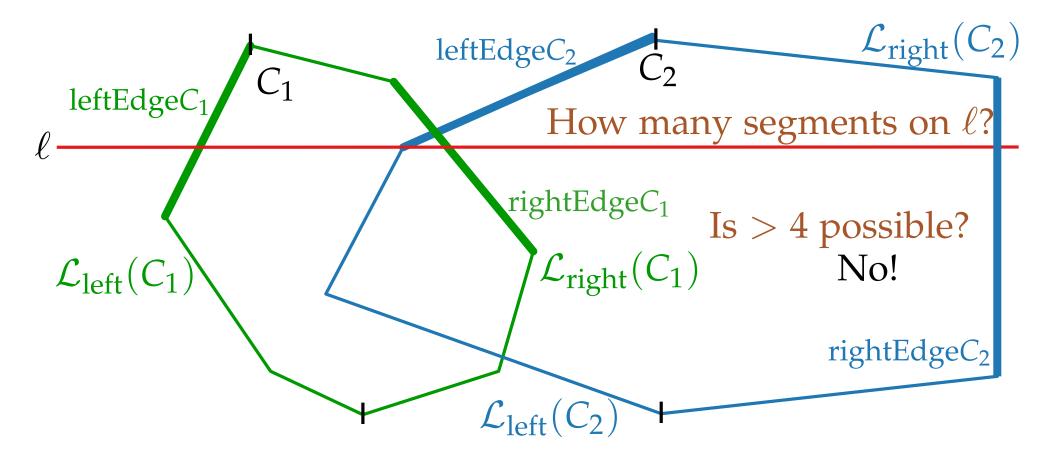
Second Approach

- compute $\bigcap H$ via divide and conquer
- walk $\partial (\cap H)$, find vertex $x \le w / cx$ maximum, O(n) time

```
IntersectHalfplanes(H)
                                     How complex can
                                     the new region be?
  if |H| = 1 then
      C \leftarrow h, where \{h\} = H
  else
      split H into sets H_1 and H_2 with |H_1|, |H_2| \approx |H|/2
      C_1 \leftarrow \text{IntersectHalfplanes}(H_1)
      C_2 \leftarrow \text{IntersectHalfplanes}(H_2)
      C \leftarrow IntersectConvexRegions(C_1, C_2)
  return C
                                                         How??
```

Running time: $T_{IH}(n) = 2T_{IH}(n/2) + T_{ICR}(n)$

Intersecting Convex Regions

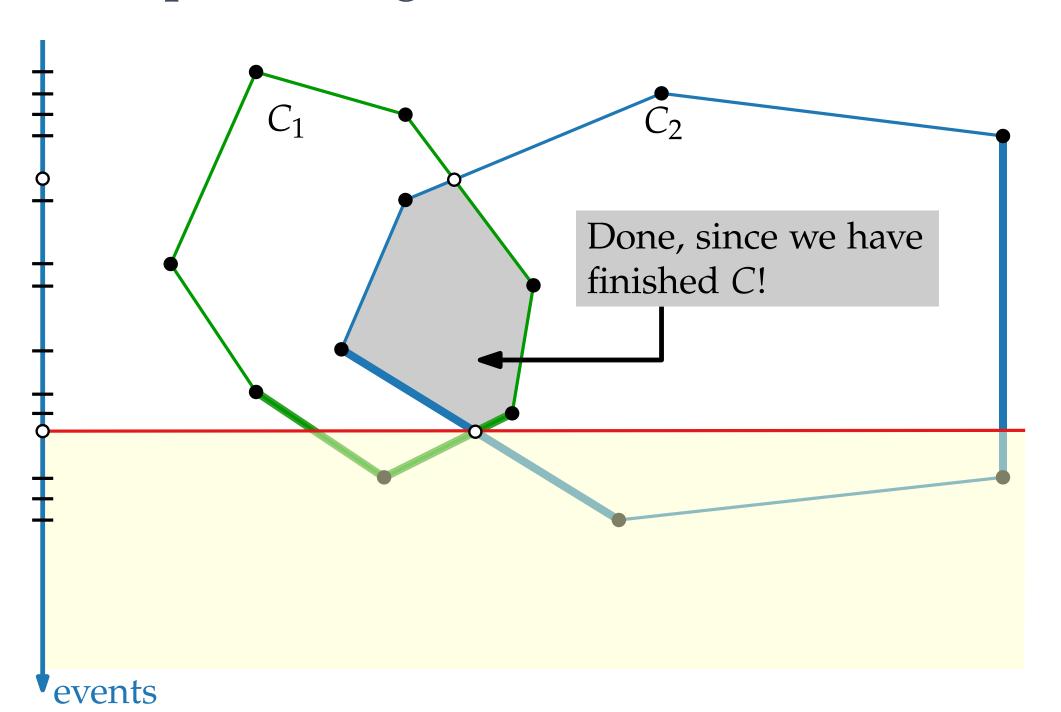


How does this help us?

→ sweep-line algorithm!

Theorem. The intersection of two convex polygonal regions can be computed in linear time.

Sweep-Line Algorithm



Data Structures

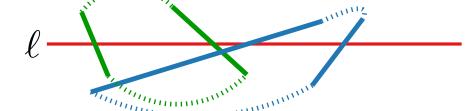
1) event (-point) queue Q

Store event pts in *sorted order* acc. to \prec

nextEvent() : either, next point (by \prec), or the intersection pt. of two active segments (below the sweep-line)

... runtime? O(1), since num. active segments ≤ 4 :)

2) (sweep-line) status \mathcal{T}



Store the segments intersected by ℓ in left-to-right order. Also, maintain the new convex hull.

Second Approach: Halfplane Intersection

Theorem. The intersection of two convex polygonal regions can be computed in linear time.

```
IntersectHalfplanes(H)

if |H| = 1 then C \leftarrow h, where \{h\} = H

else

split H into sets H_1 and H_2 with |H_1|, |H_2| \approx |H|/2

C_1 \leftarrow \text{IntersectHalfplanes}(H_1)

C_2 \leftarrow \text{IntersectHalfplanes}(H_2)

C \leftarrow \text{IntersectConvexRegions}(C_1, C_2)

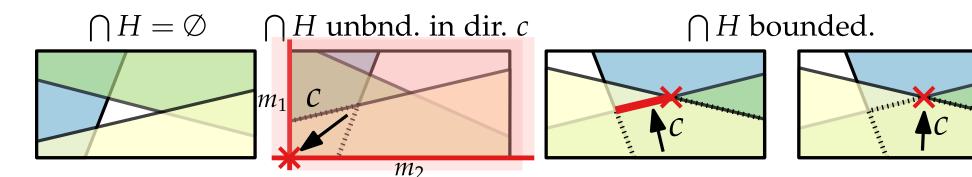
return C
```

Running time: $T_{IH}(n) = 2T_{IH}(n/2) + T_{ICR}(n)$

Corollary. The intersection of n half planes can be computed in $O(n \log n)$ time.

Can we do better?

A Small Trick: Make Solution Unique



• Add two bounding halfplanes m_1 and m_2

$$m_1 = \begin{cases} x \le M & \text{if } c_x > 0, \\ x \ge M & \text{otherwise,} \end{cases}$$
 for some sufficiently large M

$$m_2 = \begin{cases} y \le M & \text{if } c_y > 0, \\ y \ge M & \text{otherwise.} \end{cases}$$

Idea: *M* based on obj.fct. *c*. see §4.5 of CG: A&A for more on unbounded LPs.

- Take the lexicographically largest solution.
- \Rightarrow Set of solutions is either empty or a uniquely defined pt.

Incremental Approach

Idea: Don't compute $\bigcap H$, but just *one* (optimal) point! *Randomized*

```
2DBoundedLP(H, c, m_1, m_2)
   compute random permutation of H
   H_0 = \{m_1, m_2\}
   v_0 \leftarrow \text{corner of } m_1 \cap m_2
   for i \leftarrow 1 to n do
        if v_{i-1} \in h_i then
                                         \partial h_i \longrightarrow \mathbf{x}
\pi_{\partial h_i}(c)
             v_i \leftarrow v_{i-1}
                                                                              O(1)
        else
             v_i \leftarrow 1 \text{DBoundedLP}(\pi_{\partial h_i}(H_{i-1}), \pi_{\partial h_i}(c)) | O(i)
             if v_i = \text{nil then}
               return nil
                                                            w-c running time:
                                                             T(n) = \sum_{i=1}^{n} O(i) =
        H_i = H_{i-1} \cup \{h_i\} \ O(1)
                                                                     = O(n^2) :-(
   return v_n
```

Result

Theorem. The 2D bounded LP problem can be solved in O(n) expected time.

Let
$$X_i = \begin{cases} 1 & \text{if } v_{i-1} \notin h_i, \\ 0 & \text{else.} \end{cases}$$
 (indicator random variable).

Then the expected running time is

$$\mathbf{E}[T_{2d}(n)] = \mathbf{E}[\sum_{i=1}^{n} (1 - X_i) \cdot O(1) + X_i \cdot O(i)]$$

$$= O(n) + \sum_{i=1}^{n} \mathbf{E}[X_i] \cdot O(i)$$

$$= O(n) + \sum_{i=1}^{n} \mathbf{Pr}[X_i = 1] \cdot O(i) = O(n).$$

We fix the i random halfplanes in H_i .

 $Pr[X_i=1] = probability that the optimal solution changes when <math>h_i$ is added to H_{i-1} .

Proof technique: *Backward analysis!*

= probability that the optimal solution changes when h_i is removed from H_i .

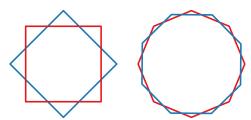
 $\leq 2/i$. This is independent of the choice of H_{i} .

Alt. for Intersecting Convex Regions



Use sweep-line alg. for map overlay (line-segment intersections)!

Running time $T_{MO}(n) = O((n + I) \log n)$,



where I = # intersection points.

here: $I \le n \longrightarrow O(n \log n)$ for ICR

Running time
$$T_{IH}(n) = 2T_{IH}(n/2) + T_{ICR}(n)$$

$$\leq 2T_{\mathrm{IH}}(n/2) + O(n\log n)$$

$$\in O(n\log^2 n)$$

As this is more general, it is unsurprisingly worse ... *

→ Better to use specialized algorithm for intersecting convex regions/polygons

^{*} it can happen sometimes that general algorithms give optimal runtimes for special cases