

Linear programming

Consider $f: \mathbb{R}^2 \rightarrow \mathbb{R}: (x,y) \mapsto c_1x + c_2y$ where $(c_1, c_2) \neq (0,0)$

& a set $H = \{h_1, \dots, h_n\}$ of halfplanes.

- Goal: find a point $(x,y) \in \cap h_i = \cap H$ at which f attains a maximal value.

- We will write $h_i : a_{i1}x + a_{i2}y \leq b_i$ for $i \in \{1, \dots, n\}$.

Geometric significance?

- f determined by vector $\vec{c} = (c_1, c_2)$

- As we move in the \vec{c} -direction

f increases: i.e. For $t > 0$

$$f(x,y) + t(c_1, c_2)$$

$$= f(x,y) + tF(c_1, c_2)$$

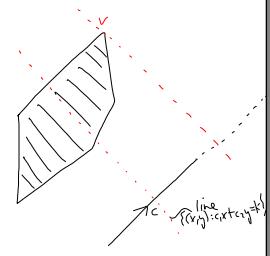
$$= f(x,y) + c_1t + c_2t > f(x,y)$$

- At lines $\{(x,y) : c_1x + c_2y = k\}$

f has constant value. These

lines are those perpendicular to \vec{c} .

- Hence f obtains maximal value at point v which is extreme in the direction of \vec{v}



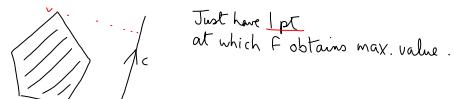
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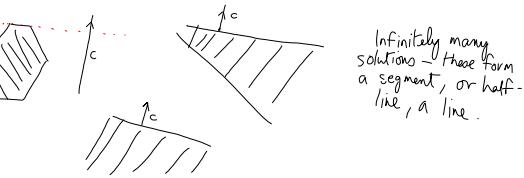
Different possibilities

1) $\cap H$ is empty. No solution - problem is infeasible.

2)

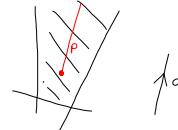
Just here LPCat which f obtains max. value.

3)



Infinitely many
solutions - these form
a segment, or half-
line, or line.

4)



The function f is unbounded on
the intersection:
in this case there exists a half-line
 f increases (P in picture).

Input to algorithm: vector $\vec{c} + \{h_1, \dots, h_n\}$
a set of halfplanes.

Output: - If problem is infeasible, provide 3 halfplanes w/ empty intersection.

- Point in intersection at which f achieves its max: If more than one point exists, choose least point with respect to a chosen lex ordering.

- If f not bounded above on intersection = provide half-line in intersection along which f is increasing.

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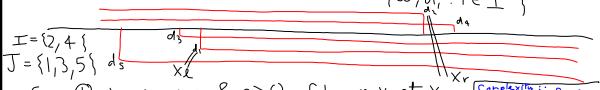
Firstly, 1-d case: $-fx = cx$ for $c \neq 0$.

- halfplanes $a_{ij}x \leq b_i$ for $a_{ij} \neq 0$ & $i=1, \dots, n$.

- let $I = \{i : a_{ij} > 0\}$, $J = \{j : a_{ij} < 0\}$

- Half-plane equations become $x \leq b_i/a_{ij}$ for $i \in I$
& $x \geq b_j/a_{ij} = d_j$ for $j \in J$.

- let $x_L = \max\{-\infty, d_j : j \in J\}$, $x_R = \min\{\infty, d_i : i \in I\}$



- Cases: ① $x_L < x_R < \infty$ & $c > 0$. f has max at x_R .

② $-\infty < x_L \leq x_R$ & $c < 0$. f has max at x_L .

③ $x_R = \infty$ (I is empty) & $c > 0$ - $[x_L, \infty)$ is half-line along which f increases.

④ $x_L = -\infty$ & $c < 0$ - then $(-\infty, x_R]$ is half-line along which f increases.

Bounded 2-d case

- In this case, we are provided 2 half-planes h_1, h_2 such that f is bounded from above on $h_1 \cap h_2$.



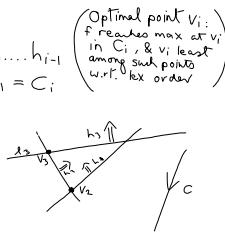
- Then $h_1 \cap h_2$ has maximum at $h_1 \cap h_2 = v$.

- If there is more than one soln - choose least one with respect to given lex ordering.

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- Algorithm is incremental : given optimal point $v_{i-1} \in C_{i-1} = h_1 \cap \dots \cap h_{i-1}$ we search for optimal point $v_i \in h_i \cap C_{i-1} = C_i$
- If $v_{i-1} \in C_i$ then $v_i = v_{i-1}$.
- Otherwise, C_i is empty or v_i lies on the boundary h_i of the half-plane h_i
- How to find v_i in this case ? let its co-ordinates be (x, y) - then $a_{ij}x + a_{iz}y = b_j$
- Assuming $a_{iz} \neq 0$ (otherwise $a_{ij} \neq 0$) we have $y = \frac{b_j - a_{ij}x}{a_{iz}}$
- We search for max value of f on this line.



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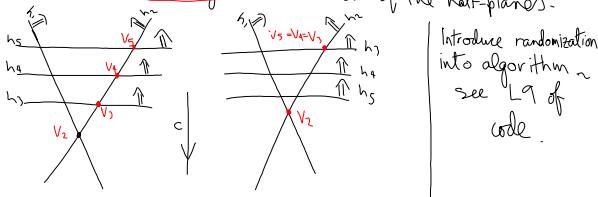
- On this line, consider F as a function of one variable :
- $g(x) = c_1x + c_2\left(\frac{b_j - a_{ij}x}{a_{iz}}\right) = \left(c_1 - c_2\frac{a_{ij}}{a_{iz}}\right)x + c_2\left(\frac{b_j}{a_{iz}}\right)$
 - Want max of g on this line - does not depend on constant c_2 . So we are seeking max of $\hat{g}(x) = \left(c_1 - c_2\frac{a_{ij}}{a_{iz}}\right)x$
 - We are looking for max of f on $l \cap C_{i+1}$, & this is given by $a_{ij}x + a_{iz}\left(\frac{b_j - a_{ij}x}{a_{iz}}\right) \leq b_j$ for $j=1, 2, \dots, l$.
 - Rewrite as $\left(a_{ij} - a_{iz}\frac{a_{ij}}{a_{iz}}\right)x \leq b_j - \frac{a_{jz}b_j}{a_{iz}}$.
 - Now we find v_i (or that C_i is empty) by solving 1-d linear program $(*, *, *)$.
 - See code - lines 7-17 (on E-learning - except line 9)

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- Running Time - If $v_{i-1} \in h_i$ constant time required to set $v_i = v_{i-1}$.
- Otherwise time to calculate v_i is linear in i - so $O(i)$

$$\text{Complexity } O(3) + O(4) + \dots + O(n) = O(3+4+\dots+n)$$

This is quite high running time, $= O(n^2)$. & depends heavily on the order of the half-planes.



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- Randomized expected time of algorithm is much lower - average time of calc. taking into account all possible orders.
- Calculation of randomized expected time : X_i a random variable defined by $X_i = \begin{cases} 1 & \text{if } v_{i-1} \notin h_i \\ 0 & \text{if } v_{i-1} \in h_i \end{cases}$
 - Time of algorithm estimated by $\sum_{i=1}^n O(i)X_i$.
 - Expected time $E(X) = \sum_{i=1}^n O(i)E(X_i)$ where $E(X_i) = \text{probability}(X_i=1) = \text{prob}(v_{i-1} \notin h_i)$.
 - In fact $\text{prob}(v_{i-1} \notin h_i) = \frac{1}{2}$.
 - Hence $E(X) = \sum_{i=1}^n O(i)\frac{1}{2} = \sum_{i=1}^n O(i) = O(n)$. Expected time is linear $O(n)$.

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- Must calculate $\text{prob}(v_{i-1} \notin h_i)$.
- Now $v_i = l_j \cap h_k$ for $j, k \leq i$ & j, k min w/ these props. $P(v_{i-1} \notin h_i) = P(j=i \text{ or } k=i)$
 - There are $i(i-1)$ choices of pairs $j, k \leq i$.
 - There are $i-1$ choices in which $j=i$.
 - So $2(i-1)$ choices in which $j \text{ or } k=i$.
 - So $\text{prob}(j=i \text{ or } k=i) = \frac{2(i-1)}{i(i-1)} = \frac{2}{i}$.
- \Rightarrow Expected randomized complexity is $O(n)$. See E-Learning for unbounded case.

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