

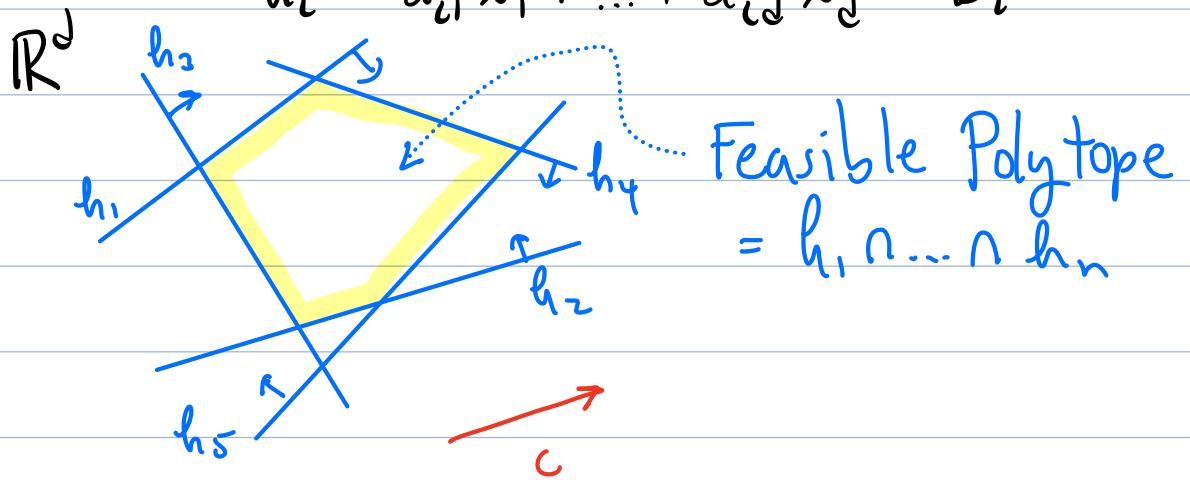
CMSC 754 - Computational Geometry

Lecture 7: Linear Programming

Linear Programming (LP):

- Fundamental optimization problem in \mathbb{R}^d
- Given a set of n linear constraints (halfspaces) $H = \{h_1, \dots, h_n\}$

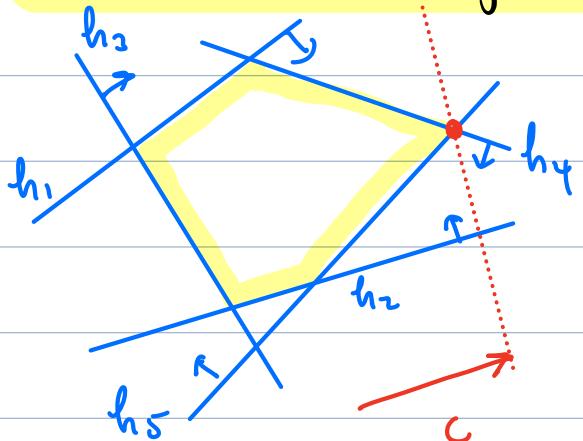
$$h_i : a_{i1}x_1 + \dots + a_{id}x_d \leq b_i$$



- Given a linear objective function

$$f(\bar{x}) = c_1x_1 + \dots + c_dx_d = c^T x$$

LP: Find the vertex of the feasible polytope that maximizes the objective function



Matrix form:

Given $c \in \mathbb{R}^d$ and $n \times d$ matrix A and $b \in \mathbb{R}^n$
find $x \in \mathbb{R}^d$ to:

maximize: $c^T x$ \leftarrow i^{th} row of A
subject to : $Ax \leq b$ corresponds to b_i

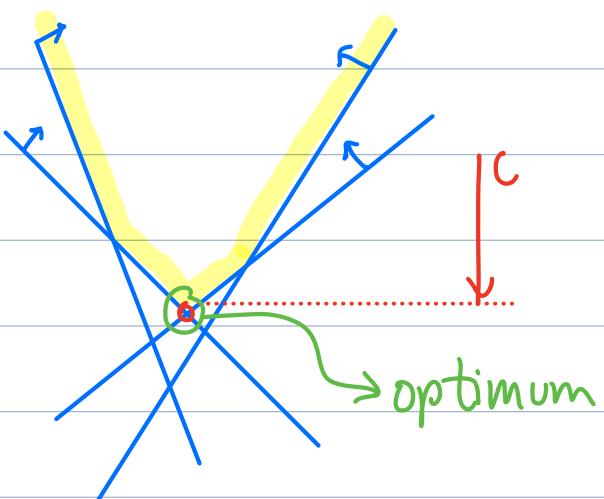
3 Possible Outcomes:

😊 Feasible: An optimal pt exists (gen'l position:
a unique vertex of feasible polytope)

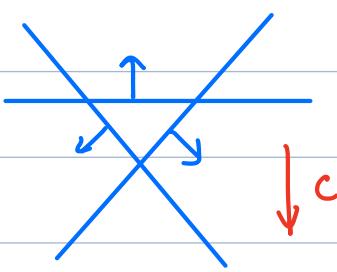
😢 Infeasible: No solution because feasible
polytope is empty

😢 Unbounded: No (finite) solution because
feasible polytope is unbounded
in direction of objective fn.

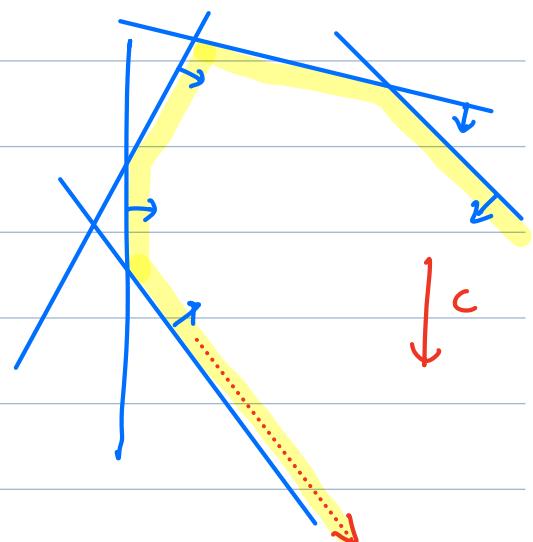
Feasible



Infeasible



Unbounded



History:

1940's : Used in operations research (Econ, Business)

Kantorovich, Dantzig, von Neuman

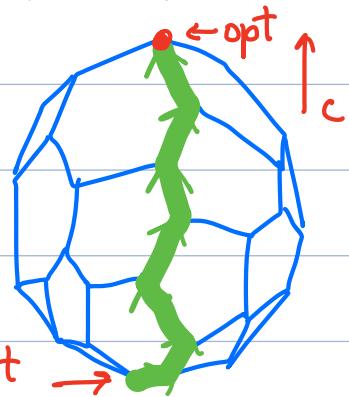
Dantzig - Simplex algorithm

(1947) - fast in practice

- exponential in worst case

↳ feasible polytope may
have $O(n^{\lfloor d/2 \rfloor})$ vertices

- Karp - Not known to be NP-hard



Khachiyan - Ellipsoid Algorithm

(1979) - (weakly) polynomial time

↳ Time depends on precision

- Compute smaller & smaller
ellipsoids containing optimum



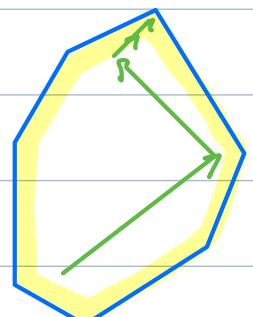
Karmarkar - Interior-Point Methods

(1984)

- Move through polytope's
interior

- (weakly) polynomial

- Practical

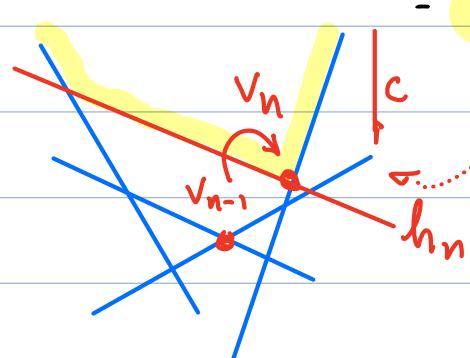
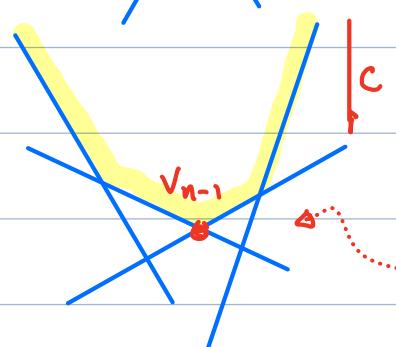
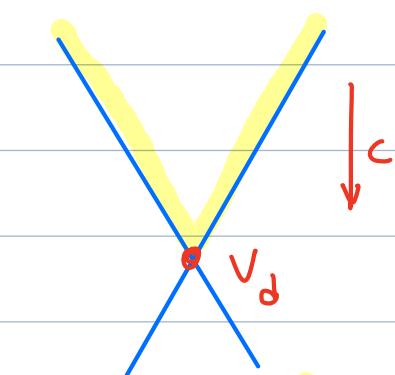


LP in constant-dimensional space

- Assume - n is large
 d is a constant
- We'll present a (randomized) algorithm with (expected) running time $O(d!n) = O(n)$

Incremental Approach:

Overview:



- Find d -halfspaces that define an initial vertex v_d (or report that LP is unbounded)

→ $O(dn)$ time (see our text)

- Remove halfspace h_n and recursively compute LP on $n-1$ halfspaces h_1, \dots, h_{n-1}

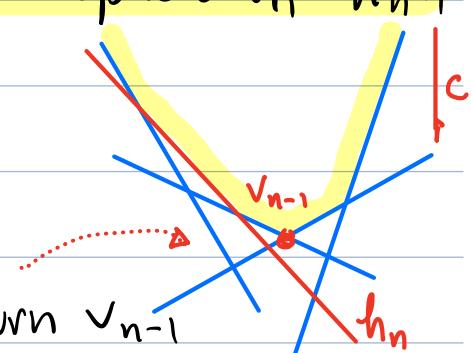
If infeasible → return

else let v_{n-1} be opt

- Add back h_n

- If ($v_{n-1} \in h_n$) return v_{n-1}

- else ...



How to update opt. vertex?

Lemma: If $v_{n-1} \notin h_n$ then new opt vertex (v_n) lies on the hyperplane bounding h_n .

Proof: Let h_n be hyperplane bounding h_n . Assume c directed downwards.

v_{n-1} - not feasible \Rightarrow below h_n

v_n - if not on $h_n \Rightarrow$ above h_n

Let $p = h_n \cap \overline{v_{n-1} v_n}$

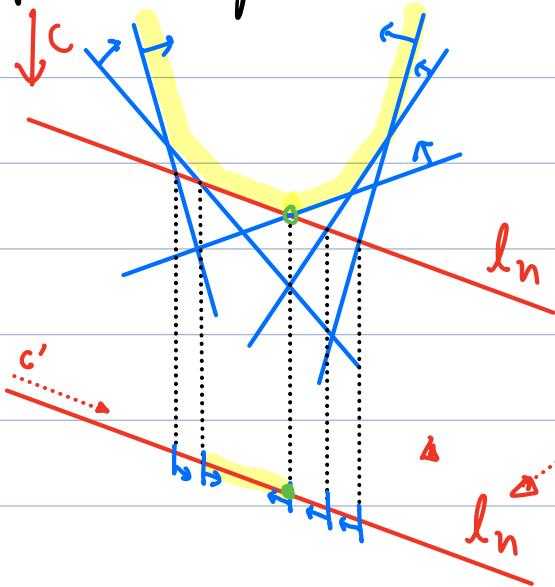
By convexity, $p \in$ feasible polytope

By linearity, obj. function gets progressively worse from $v_{n-1} \rightarrow v_n$

$\Rightarrow p$ is better solution than v_n

* contradiction!

How to update?



- Intersect h_1, \dots, h_{n-1}

with $h_n - O(d \cdot n)$

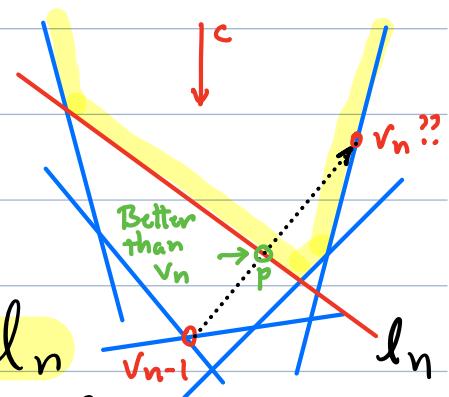
- This yields a $(d-1)$ -dim polytope

- Project c onto $h_n \rightarrow c'$

- Solve this $(d-1)$ -dim LP

recursively (If $d=1$, solve by brute force $O(n)$)

(See latex notes for details)



Running time? Pretty bad - $\mathcal{O}(n^d)$

- Let $\bar{W}_d(n)$ be worst-case complexity for n halfspaces in dim d

- Recurrence:

$$\bar{W}_d(n) = \bar{W}_d(n-1) + d + dn + \bar{W}_{d-1}(n-1)$$

solve LP on
 h_1, \dots, h_{n-1}

Test
 $v_{n-1} \in h_n$

Project
onto l_n

Solve LP on
 l_n

Claim: $\bar{W}_d(n) = \mathcal{O}(n^d)$

Sketch: Very similar recurrence:

$$\bar{W}'_d(n) = \bar{W}'_d(n-1) + \bar{W}'_{d-1}(n)$$

Note similarity with binomial coeffs:

$$\binom{n}{d} = \binom{n-1}{d} + \binom{n}{d-1}$$

It is well known that $\binom{n}{d} = \mathcal{O}(n^d)$

Applies to \bar{W}' as well.

How to fix this?

Easy! Randomize the choice of h_n

Why?

$$\bar{W}_d(n) = \bar{W}_d(n-1) + d + dn + \bar{W}_{d-1}(n-1)$$

This solves
to $\mathcal{O}(n)$

Only applies if
 $v_{n-1} \notin h_n$

This rarely happens!

Randomized Incremental Algorithm

Input: $H = \{h_1, \dots, h_n\}$ constraint halfspaces in \mathbb{R}^d

$c \in \mathbb{R}^d$ objective vector

Output: Optimum vertex v or error {unbounded or infeasible}

(1) If ($d=1$) solve LP by brute force - $O(n)$

(2) Find initial subset $\{h_1, \dots, h_d\}$ that provide initial optimum v_d (or return "unbounded")
- $O(d \cdot n)$ (see text)

(3) Randomly select halfspace from $\{h_{d+1}, \dots, h_n\}$

- call it h_n . Recursively solve LP on remaining $n-1$ halfspaces → Let v_{n-1} be result

(4) If ($v_{n-1} \in h_n$) return v_{n-1} → $O(d)$

(5) else, project $\{h_1, \dots, h_{n-1}\} + c$ onto h_n ,
the bounding hyperplane for h_n .

Solve recursively, letting v_n be result. Return v_n

Expected Case Running Time:

- The running depends on the (random) choice of h_n

- Let $T_d(n)$ be the expected-case running time, over all choices of h_n .

- Let p_n = probability that $v_{n-1} \notin h_n$

- To simplify, assume all halfspaces chosen randomly (h_1, \dots, h_d aren't)

Recurrence:

$$T_d(n) = \begin{cases} 1 & \text{if } n=1 \\ n & \text{if } d=1 \\ T_d(n-1) + d + p_n(dn + T_{d-1}(n-1)) & \text{o.w.} \end{cases}$$

(3) Recursively compute v_{n-1}

(4) test if $v_{n-1} \in h_n$

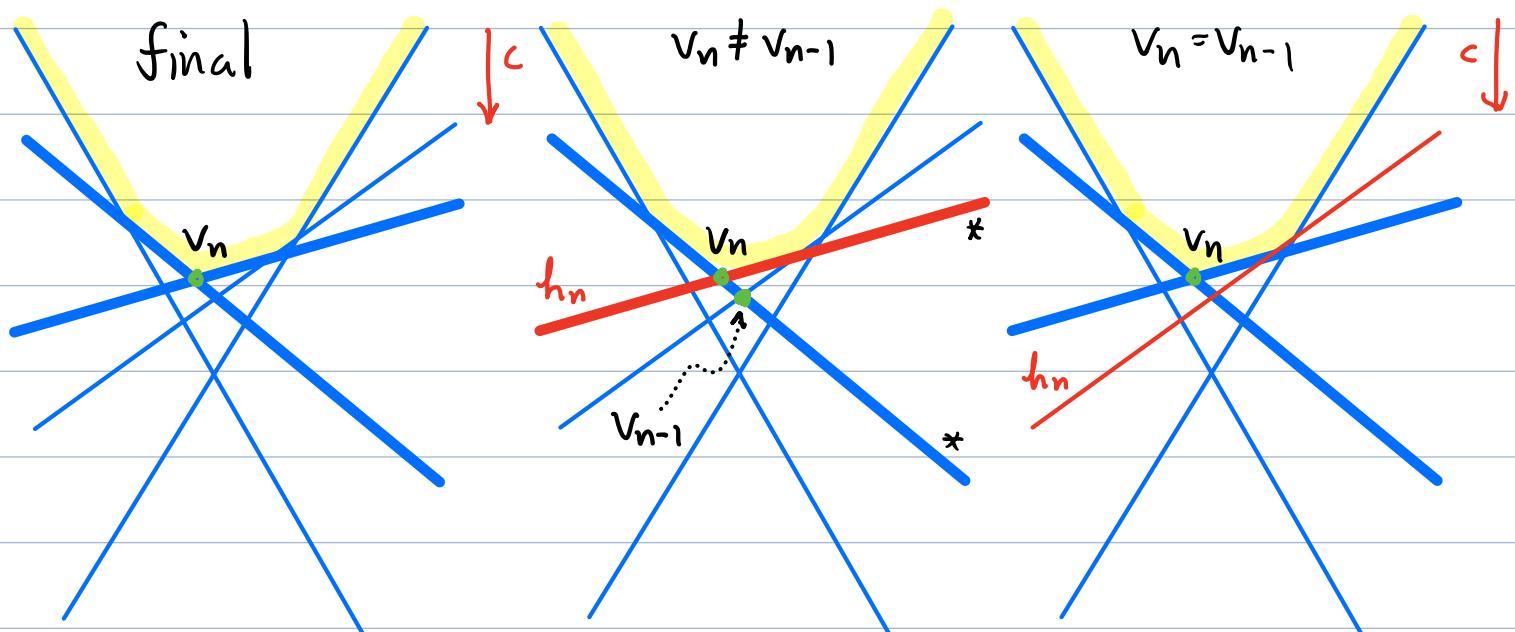
if not

(5) project h_1, \dots, h_{n-1} onto l_n

(5) solve $d-1$ dim LP on projections

What is p_n ? Backwards Analysis

- Let's consider the final configuration and ask - which halfspace came last and how does its choice affect things?



Obs: The optimum is determined by d halfspaces (assuming gen'l position)

- If h_n is any of these, $v_{n-1} \notin h_n + v_n \neq v_{n-1}$ 😞
- Otherwise, $v_{n-1} \in h_n + v_n = v_{n-1}$ 😊

$$\Rightarrow p_n = d/n$$

If $n \gg d$, p_n very small
+ bad case unlikely

Why is it called "backwards"?

- We consider final config. and look backwards to our last random choice

Lemma: $T_d(n) \leq \gamma_d d! n$, where γ_d is a constant depending on dimension

Proof: Induction on $n+d$

$$T_d(n) = T_d(n-1) + d + p_n (d \cdot n + T_{d-1}(n))$$

by I.H. $\leq \gamma_d \cdot d! (n-1) + d + \frac{d}{n} (d \cdot n + \gamma_{d-1} (d-1)! n)$
+ def of p_n

$$= \gamma_d d! (n-1) + d + (d^2 + \gamma_{d-1} d!)$$

$$= \gamma_d d! n + (d + d^2 + \gamma_{d-1} d! - \gamma_d d!)$$

want:

$$\leq \gamma_d d! n$$

Suffices to select γ_d such that

$$d + d^2 + \gamma_{d-1} d! - \gamma_d d! \leq 0$$


$$\Leftrightarrow d! \gamma_d \geq d + d^2 + \gamma_{d-1} d!$$

We can satisfy this by setting:

$$\gamma_1 \leftarrow 1$$

$$\gamma_d \leftarrow \frac{d+d^2}{d!} + \gamma_{d-1}$$

$\Rightarrow \gamma_d$ is a constant
depending on dim



Summary:

- Randomized algorithm for LP
- Expected run time of LP is $O(d! n) = O(n)$
(since we assume d is constant)
- Variation depends on random choices, not input
- (Seidel) Prob of running slower extremely small