

# Algorithms - Spring '25

LP: Simplex

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# Recap

- Oral grading this week  
(No office hours for me today)
- HW 6 graded, HW7 coming soon
- Practice final: any questions on general topics/format?
- Wed: practice problems during class

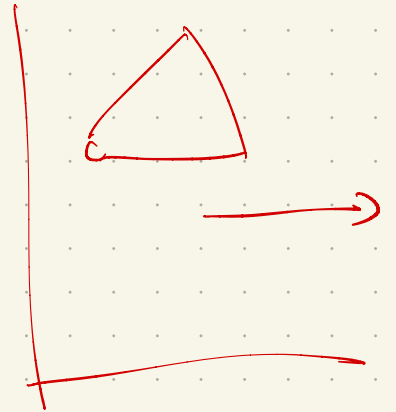
# The algorithm: Simplex (Dantzig 1947)

Assumes canonical form:

$$\text{maximize } \sum_{j=1}^d c_j x_j$$

$$\text{subject to } \sum_{j=1}^d a_{ij} x_j \leq b_i \quad \text{for each } i = 1 \dots n$$

$$x_j \geq 0 \quad \text{for each } j = 1 \dots d$$



So.

- no min
- only  $\leq$
- $\geq 0$  for all variables
- fast in practice, but exponential in worst case

Klee-Minty, 1973: some feasible polytopes have  $O(n^{\lfloor d/2 \rfloor})$  vertices

# Algorithm (Simplex):

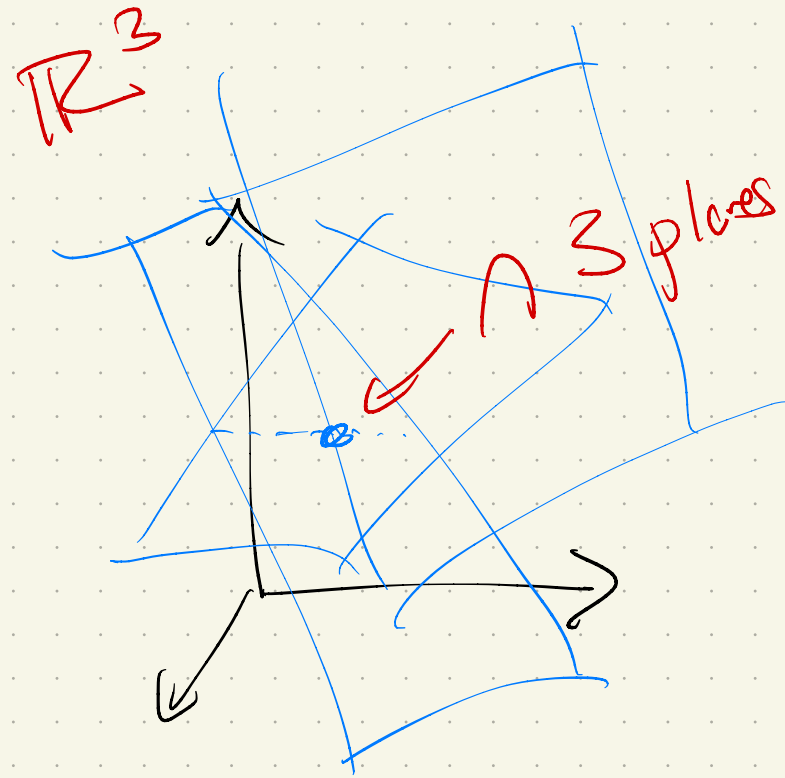
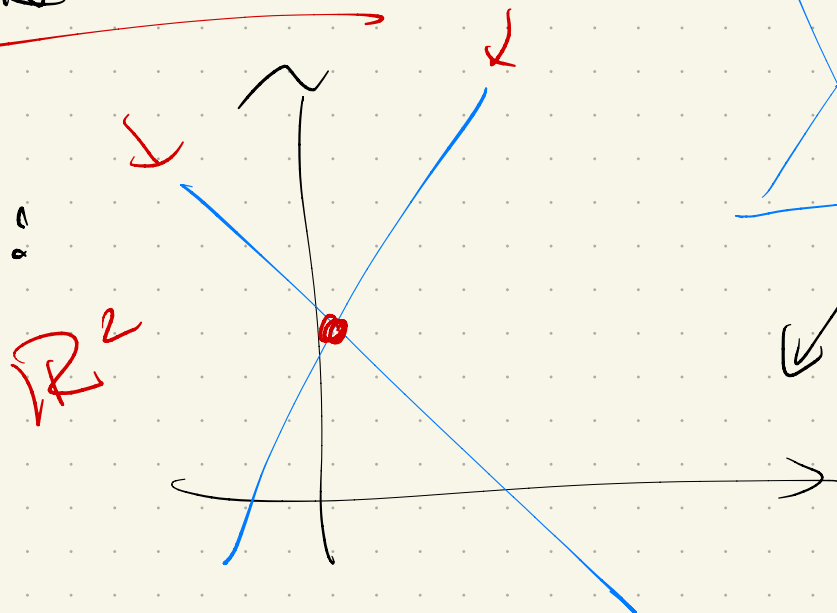
Take any vertex  $v$  in feasible region  
while some neighbor  $v'$  is better

$$v \leftarrow v'$$

Details lacking!

Step 1: find a vertex

take  $d$   
hyperplanes:

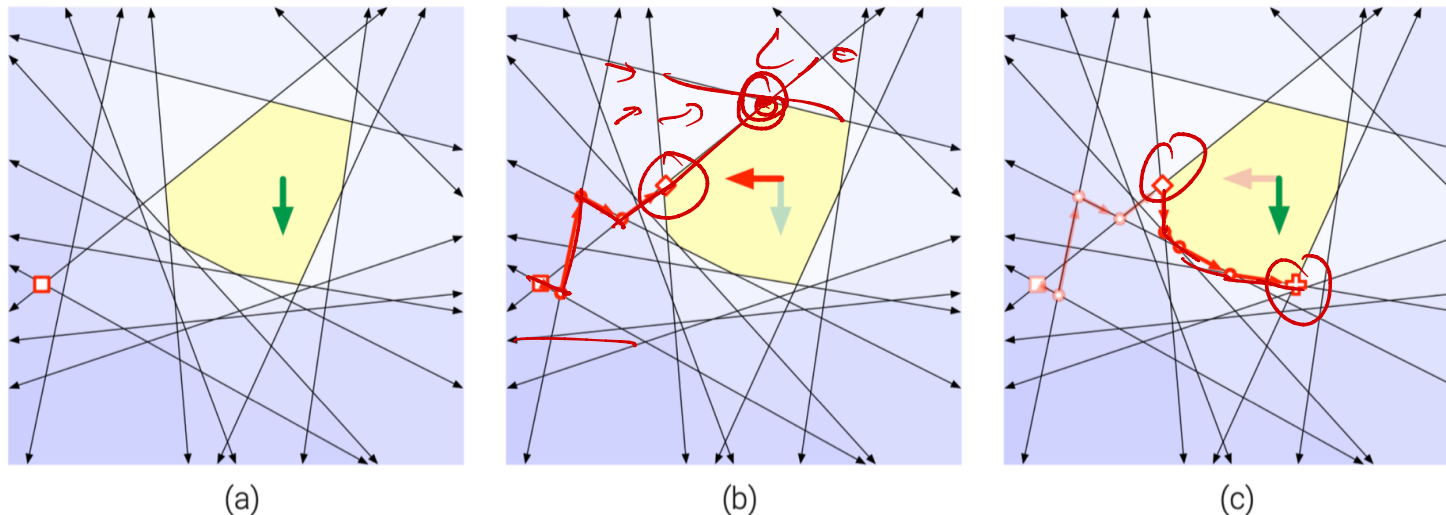




Any  $d$  hyperplanes give a vertex:  
Loop through  $m-d$  others!

Either feasible for each  
Or Not

↳ use this new hyperplane

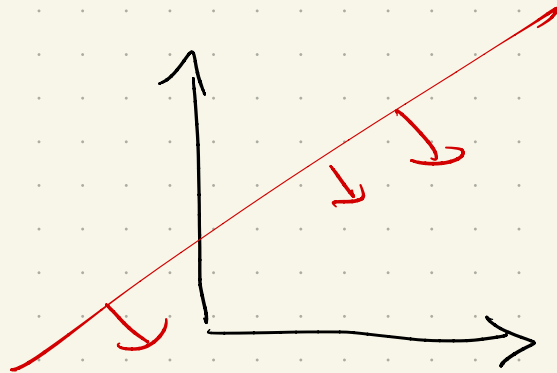


**Figure 1.2.** Primal simplex with dual initialization: (a) Choose any basis. (b) Rotate the objective to make the basis locally optimal, and pivot "up" to a feasible basis. (c) Pivot down to the optimum basis for the original objective.

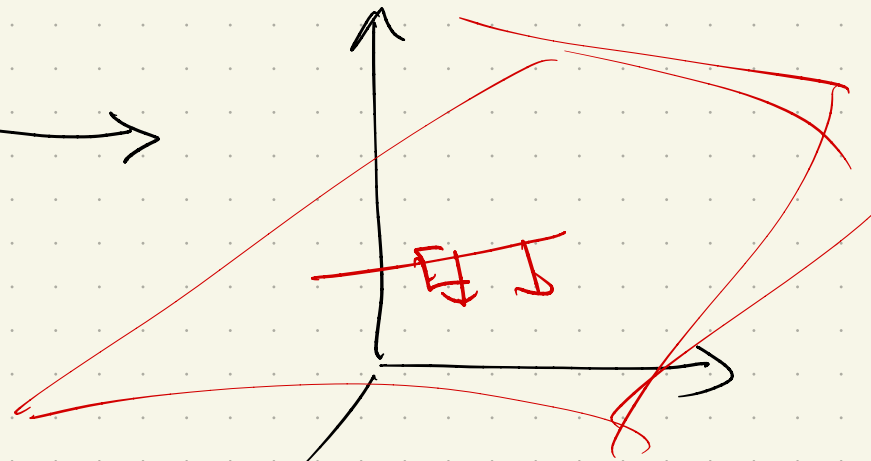
Let's go over that more carefully:

Each LP equality or inequality describes a hyperplane in  $\mathbb{R}^d$ .

$$2d: ax + by \leq c$$



$$3d: ax + by + cz \leq d$$



$d$  variables

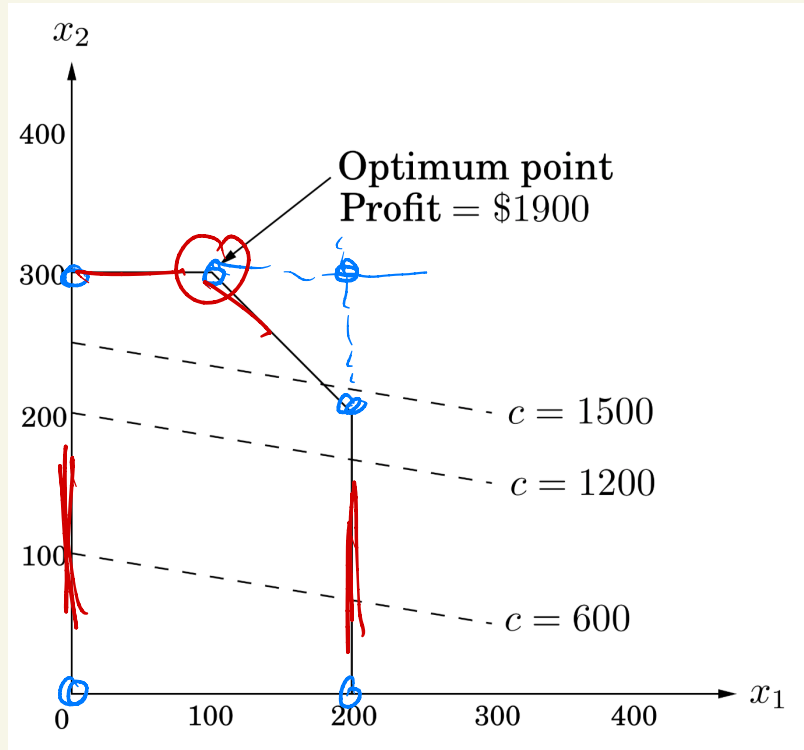
$$\mathbb{R}^d: a_1x_1 + a_2x_2 + \dots + a_dx_d \leq c$$

## Vertices:

These happen when  $> d$  hyperplanes meet in  $\mathbb{R}^d$

In  $\mathbb{R}^2$

eqn 1  
eqn 2  
→ solve for point



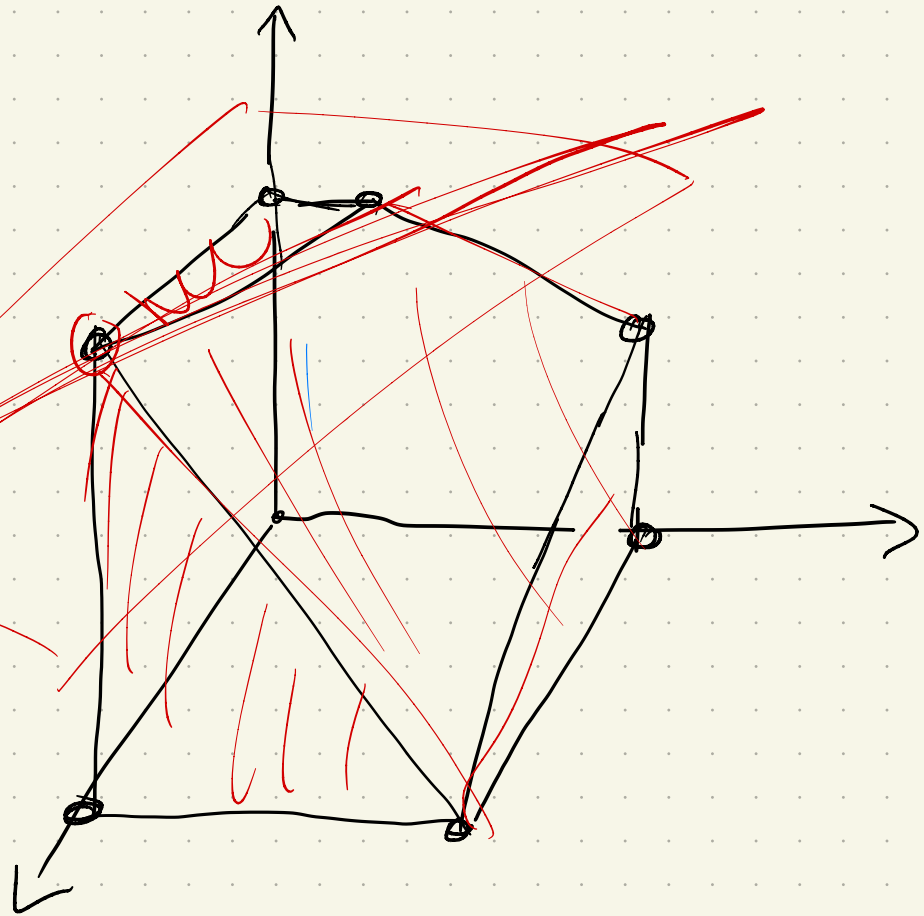
Note:

Not every pair will meet!

In  $\mathbb{R}^3$

Any 2 intersect  
in a line.

Any 3 intersect  
in a vertex  
(assuming not  
parallel)

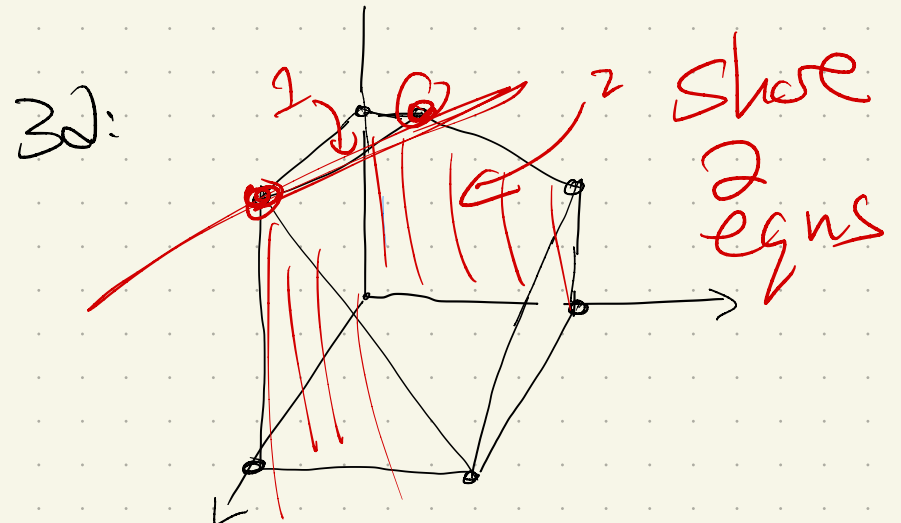
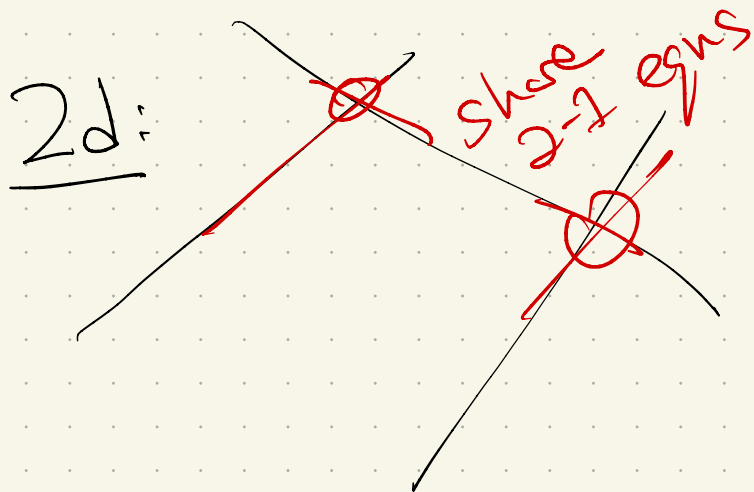


In  $\mathbb{R}^d$ :  $d$  equations  
 $d$  variables }  $\Rightarrow$  one point

Dfn: Pick a subset of inequalities

If there is a unique point that satisfies  
all with equality, & it is feasible  
→ this is a vertex of the solution

Since each vertex is specified by  
d equations, we call any two  
that share  $(d-1)$  of them neighbors:



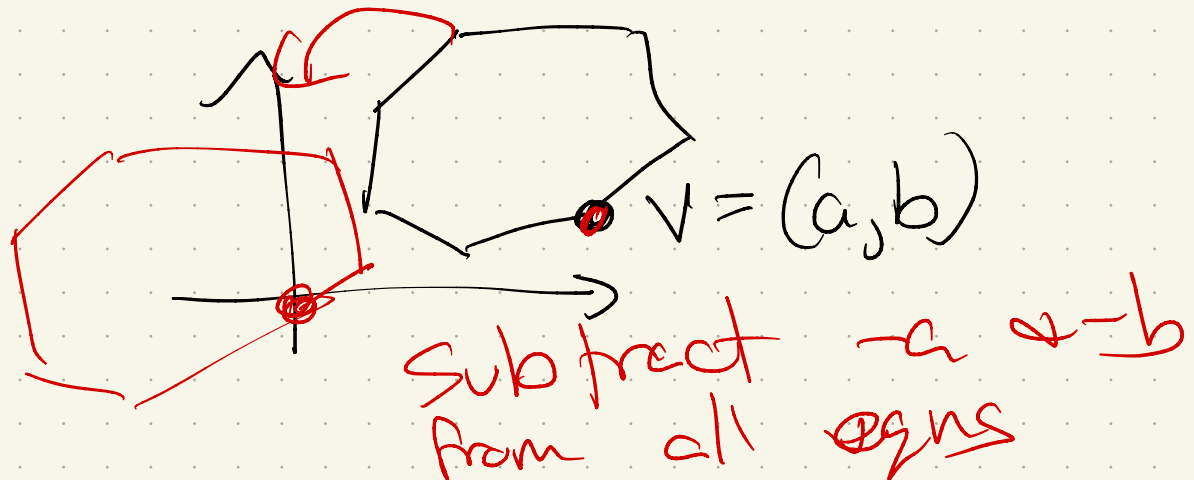
# Simplex algorithm:

In each stage, 2 tasks:

- ① Check if current vertex is optimal
- ② If not choose a neighbor that gives a better score under the objective function

Both are easy if  $v = \vec{0}$  (the origin)  
(see next slide)

If not at  $\vec{0}$ :  
translate



LP:  $\max c^T x$   
s.t.  $A\vec{x} \leq \vec{b}$

$x_i \geq 0 \quad \forall i$

Note:  $\vec{x} \in \mathbb{R}^d$ , so  $x = (x_1, \dots, x_d)$

Now,  $\vec{0}$  is always a vertex - why?  
 $\rightarrow$  these  $d$  eqns intersect at  $\vec{0}$

Optimal only if all  $c_i$ 's are negative

Conversely:  $\tau \rightarrow \text{obj}$  then is  $\downarrow$

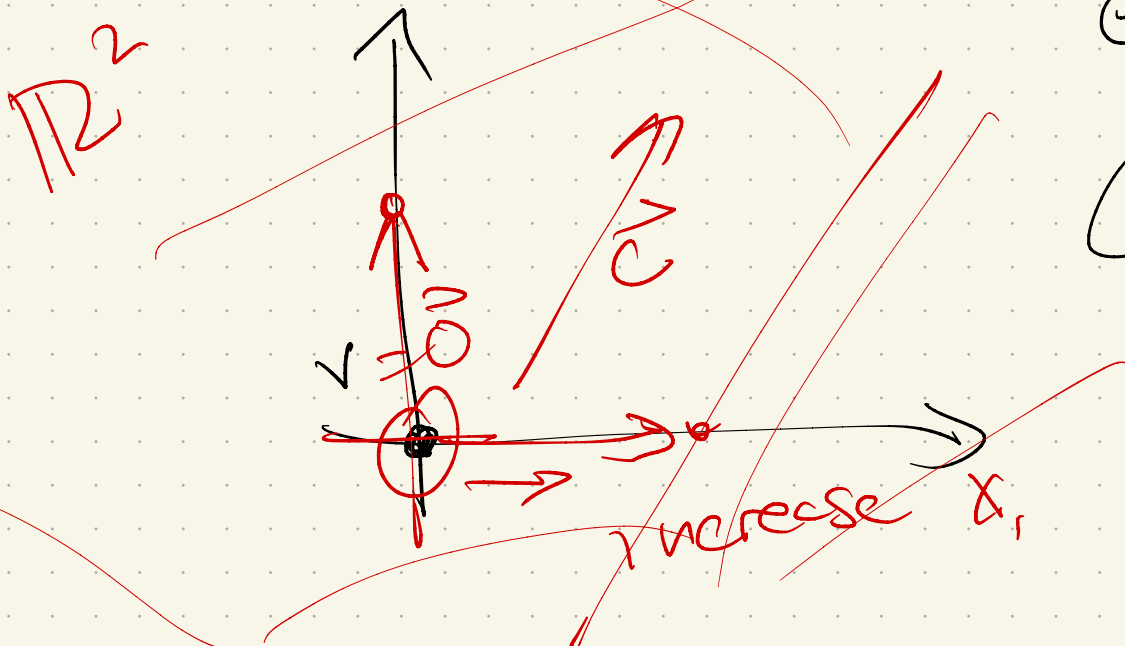
If any  $c_i > 0$ , we can increase the obj. function  $C^T \vec{x}$

How?

So: pick one & increase 1

How much? until stuck on some edge

either  $c_1$  or  $c_2$  is positive  
 ↳ when do we get stuck?



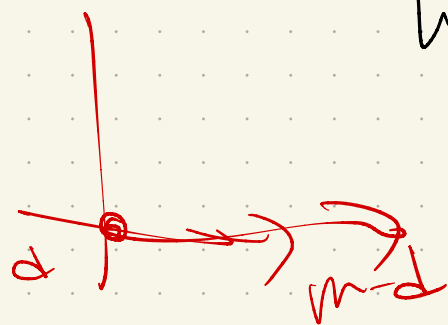


Runtime:

Consider a vertex  $\vec{v} \in \mathbb{R}^d$  with  $m$   
inequalities &  ~~$m$~~  <sub>$d$</sub>  variables

How many possible neighbors?

each removes 1 eqn & replaces  
with another



$\leq$

$$d \cdot (m-d)$$

*other eqns*

Checking if it is a neighbor & is  
feasible: basically dot product &  
Gaussian elimination.

So: each iteration is  
 $\leq \underbrace{d \cdot (m-d)}_{\text{time for G.E.}} \cdot \underbrace{\left[ \text{time for G.E.} \right]}_{\text{time for G.E.}}$

Can improve slightly:

- just need one  $c_i > 0$  + rescaling  
to 0 is easy.

So can get to  $O(m \cdot n)$   
per nbr we find.

How many times do we need  
to find a nbr?

How many iterations?

$\left\{ \begin{array}{l} m+d \text{ inequalities} \\ \text{any } d \text{ give a vertex} \end{array} \right\} \Rightarrow \binom{m+d}{d} \sim O(m^d)$

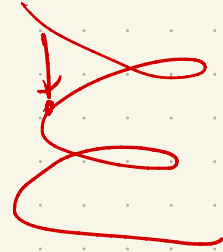
So exponential in worst case.

(Remember, for a while people thought this might be NP-Hard)

Klee & Minty gave examples in the ~~50's~~<sup>70's</sup> that actually take exponential time.

Can we avoid this by choosing the "best" neighbor in our update?

No ideal way!



Many proposed, but for almost every one, there is some input polyhedron that needs an exponential number of pivots.

Ellipsoid algorithm, Khachiyan 1979

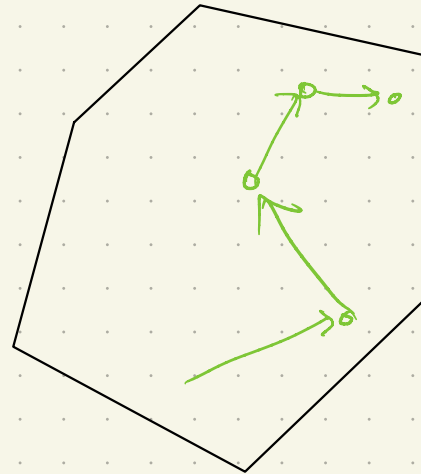
- (weakly) polynomial time  
↳ dependent of precision



- high level idea: compute smaller & smaller ellipses which contain solution

Interior point Methods, Karmarkar 1984

- Move through polytope's interior:
- Still weakly polynomial
- but - practical



More recent:

- Matrix Multiply time (in 2019!)

RESEARCH-ARTICLE

## Solving linear programs in the current matrix multiplication time

Authors:  [Michael B. Cohen](#),  [Yin Tat Lee](#),  [Zhao Song](#) | [Authors Info & Claims](#)

STOC 2019: Proceedings of the 51st Annual ACM SIGACT Symposium on Theory of Computing • Pages 938 - 942  
<https://doi.org/10.1145/3313276.3316303>

Published: 23 June 2019 [Publication History](#)



### Abstract

This paper shows how to solve linear programs of the form  $\min_{Ax=b, x \geq 0} c^T x$  with  $n$  variables in time  $O^*$   
 $((n^\omega + n^{2.5-\alpha/2} + n^{2+1/6}) \log(n/\delta))$  where  $\omega$  is the exponent of matrix multiplication,  $\alpha$  is the dual exponent of matrix  
multiplication, and  $\delta$  is the relative accuracy. For the current value of  $\omega \sim 2.37$  and  $\alpha \sim 0.31$ , our algorithm takes  $O^*(n^\omega \log(n/\delta))$  time. When  $\omega = 2$ , our algorithm takes  $O^*(n^{2+1/6} \log(n/\delta))$  time.

Our algorithm utilizes several new concepts that we believe may be of independent interest:

- (1) We define a stochastic central path method.
- (2) We show how to maintain a projection matrix

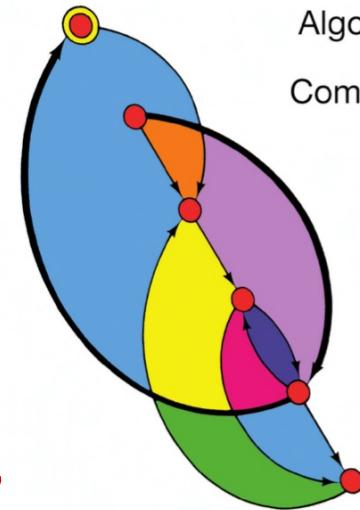
$$\sqrt{W} A^T (A W A^T)^{-1} A \sqrt{W}$$

in sub-quadratic time under  $\ell_2$  multiplicative changes in the diagonal matrix  $W$ .

This is a large & active  
area of study:

## COMBINATORIAL OPTIMIZATION

Algorithms  
and  
Complexity



Christos H. Papadimitriou  
Kenneth Steiglitz

Now:

5 minutes to do CIFS,  
if not already done!

thanks! →

