

## Linear programming.

- define
- assumptions:
  - nonempty, bounded polyhedron
  - minimizing  $x_1$
  - unique minimum, at a vertex
  - exactly  $d$  constraints per vertex
- definitions:
  - hyperplanes  $H$
  - **basis**  $B(H)$  of hyperplanes that define optimum
  - optimum value  $O(H)$
- Simplex
  - exhaustive polytope search:
  - walks on vertices
  - runs in  $O(n^{\lceil d/2 \rceil})$  time in theory
  - often great in practice
- polytime algorithms exist (ellipsoid)
- but bit-dependent (weakly polynomial)!
- OPEN: strongly polynomial LP
- goal today: polynomial algorithms for small  $d$

### Random sampling algorithm

- Goal: find  $B(H)$
- Plan: random sample
  - solve random subproblem
  - keep only violating constraints  $V$
  - recurse on leftover
- problem: violators may not contain all of  $B(H)$
- bf BUT, contain **some** of  $B(H)$ 
  - opt of sample better than opt of whole

- but any point feasible for  $B(H)$  no better than  $O(H)$
- so current opt not feasible for  $B(H)$
- so some  $B(H)$  violated
- revised plan:
  - random sample
  - discard useless planes, add violators to “active set”
  - repeat sample on whole problem while keeping active set
  - claim: add one  $B(H)$  per iteration
- Algorithm **SampLP**:
  - set  $S$  of “active” hyperplanes.
  - if  $n < 9d^2$  do simplex ( $d^{d/2+O(1)}$ )
  - pick  $R \subseteq H - S$  of size  $d\sqrt{n}$
  - $x \leftarrow \mathbf{SampLP}(R \cup S)$
  - $V \leftarrow$  hyperplanes of  $H$  that violate  $x$
  - if  $V \leq 2\sqrt{n}$ , add to  $S$
- Runtime analysis:
  - mean size of  $V$  at most  $\sqrt{n}$
  - each iteration adds to  $S$  with prob.  $1/2$ .
  - each successful iteration adds a  $B(H)$  to  $S$
  - deduce expect  $2d$  iterations.
  - $O(dn)$  per phase needed to check violating constraints:  $O(d^2n)$  total
  - recursion size at most  $2d\sqrt{n}$

$$T(n) \leq 2dT(2d\sqrt{n}) + O(d^2n) = O(d^2n \log n) + (\log n)^{O(\log d)}$$

(Note valid use of linearity of expectation)

Must prove claim, that mean  $V \leq \sqrt{n}$ .

- Lemma:
  - suppose  $|H - S| = m$ .
  - sample  $R$  of size  $r$  from  $H - S$
  - then expected violators  $d(m - r - 1)/(r - d)$

- **book broken: only works for empty  $S$**

- Let  $C_H$  be set of optima of subsets  $T \cup S$ ,  $T \subseteq H$
- Let  $C_R$  be set of optima of subsets  $T \cup S$ ,  $T \subseteq R$
- note  $C_R \subseteq C_H$ , and  $O(R \cup S)$  is only point violating no constraints of  $R$
- Let  $v_x$  be number of constraints in  $H$  violated by  $x \in C_H$ ,
- Let  $i_x$  indicate  $x = OPT(R \cup S)$

$$\begin{aligned} E[|V|] &= E\left[\sum v_x i_x\right] \\ &= \sum v_x \Pr[i_x] \end{aligned}$$

- decide  $\Pr[v_x]$ 
  - $\binom{m}{r}$  equally likely subsets.
  - how many have optimum  $x$ ?
  - let  $q_x$  be number of planes defining  $x$  **not** already in  $S$
  - must choose  $q_x$  planes to define  $x$
  - all others choices must avoid planes violating  $x$ . prob.

$$\begin{aligned} \binom{m - v_x - q_x}{r - q_x} / \binom{m}{r} &= \frac{(m - v_x - q_x) - (r - q_x) + 1}{r - q_x} \binom{m - v_x - q_x}{r - q_x - 1} / \binom{m}{r} \\ &\leq \frac{(m - r + 1)}{r - d} \binom{m - v_x - q_x}{r - q_x - 1} / \binom{m}{r} \end{aligned}$$

- deduce

$$E[V] \leq \frac{m - r + 1}{r - d} \sum v_x \binom{m - v_x - q_x}{r - q_x - 1} / \binom{m}{r}$$

- summand is prob that  $x$  is a point that violates exactly one constraint in  $r$ .
  - \* must pick  $q_x$  constraints defining  $x$
  - \* must pick  $r - q_x - 1$  constraints from  $m - v_x - q_x$  nonviolators
  - \* must pick one of  $v_x$  violators
- therefore, sum is expected number of points that violate exactly one constraint in  $R$ .
- but this is only  $d$  (one for each constraint in basis of  $R$ )

Result:

- saw sampling LP that ran in time  $O((\log n)^{O(\log d)} + d^2 n \log n + d^{O(d)})$
- key idea: if pick  $r$  random hyperplanes and solve, expect only  $dm/r$  violating hyperplanes.

## Iterative Reweighting

Get rid of recursion and highest order term.

- idea: be “softer” regarding mistakes
- plane in  $V$  gives “evidence” it’s in  $B(H)$
- Algorithm:
  - give each plane weight one
  - pick  $9d^2$  planes with prob. proportional to weights
  - find optimum of  $R$
  - find violators of  $R$
  - if

$$\sum_{h \in V} w_h \leq (2 \sum_{h \in H} w_h) / (9d - 1)$$

then double violator weights

- repeat till no violators
- Analysis
  - show weight of basis grows till rest is negligible.
  - claim  $O(d \log n)$  iterations suffice.
  - claim iter successful with prob.  $1/2$
  - deduce runtime  $O(d^2 n \log n) + d^{d/2+O(1)} \log n$ .
  - proof of claim:
    - \* after each iter, double weight of some basis element
    - \* after  $kd$  iterations, basis weight at least  $d2^k$
    - \* total weight increase at most  $(1 + 2/(9d - 1))^{kd} \leq n \exp(2kd/(9d - 1))$
  - after  $d \log n$  iterations, done.
- so runtime  $O(d^2 n \log n) + d^{O(d)} \log n$
- Can improve to linear in  $n$

## Randomized incremental algorithm

$$T(n) \leq T(n - 1, d) + \frac{d}{n}(O(dn) + T(n - 1, d - 1)) = O(d!n)$$

Incomparable to prior bound.

Improvement to Seidel:

- Silly to discard previous info on recursion

- tested basis  $B$ , violated by  $H$
- start from basis of  $B \cup \{h\}$
- Intuition: forms good starting point for recursive call
- “hidden dimension” is how many of true basis hyperplanes are in current bases
- show hidden dimension rises quickly
- improves bound to  $O(d^4 2^d N)$  (see book)

Followups:

- Kalai achieved  $n^{O(\sqrt{d \log d})}$  (subexponential)
- led to more careful analysis above:  $nd^{\sqrt{d \log n}}$
- combined with above to  $O(d^2 n + b^{\sqrt{d \log d} \log n})$

Is polynomial possible?

- these are all simplex algorithms
- cannot do better than diameter of graph
- Kalai and Kleitman proved  $n^{2+\log d}$
- must better than best algs, but still not poly

## 1 Voronoi Diagram

Goal: find nearest athena terminal to query point.

Definitions:

- point set  $p$
- $V(p_i)$  is space closer to  $p_i$  than anything else
- for two points,  $V(P)$  is bisecting line
- For 3 points, creates a new “voronoi” point
- And for many points,  $V(p_i)$  is intersection of halfplanes, so a convex polyhedron
- And nonempty of course.
- but might be infinite
- Given VD, can find nearest neighbor view planar point location:
- $O(\log n)$  using persistent trees

Space complexity:

- VD is a **planar graph**: no two voronoi edges cross (if count voronoi points)
- add one point at infinity to make it a proper graph with ends
- Euler's formula:  $n_v - n_e + n_f = 2$
- ( $n_v$  is voronoi points, not original ones)
- But  $n_f = n$
- Also, every voronoi point has degree at least 3 while every edge has two endpoints.
- Thus,  $2n_e \geq 3(n_v + 1)$
- rewrite  $2(n + n_v - 2) \geq 3(n_v + 1)$
- So  $n - 2 \geq (n_v + 3)/2$ , ie  $n_v \leq 2n - 7$
- Gives  $n_e \leq 3n - 6$

Summary:  $V(P)$  has linear space and  $O(\log n)$  query time.  
Which voronoi points and lines survive?

- if no other point inside circle containing them, then survive

## Delaunay Triangulation

For interpolation

- Given values at set of points
- interpolate elsewhere by convex combination
- eg, topographical map with heights at given points.

Goal: no skinny triangles

- Consider 4 points in convex
- two triangulations
- one makes fatter triangles
- it's the one with no points inside those triangles
- Delaunay triangles: triples with no points inside circles

Voronoi and Delunay

- Define planar dual graph
- argue based on contained circles

## Construction

Several methods

- Voronoi is projection of convex hull of lift
- Or, build Delaunay, take dual

To build Delaunay:

- Find “illegal edge”, flip

Incremental construction

- Insert point (inside some triangle)
- draw 3 lines
- flip illegal edges till stable
- Claim: Illegal edges only at changes, so can propagate from insertion
- Claim: All flips produce edges incident on new point, which are Delaunay

Analysis:

- Each flip takes constant time, so proportional to number of flips
- So proportional to final number of edges on inserted point
- RIC. Average degree constant
- So flip work per insert constant
- So  $O(n)$  flip work

Detail:

- Need to know which triangle point goes in
- Use point location like TD
- When destroy triangles, point their (leaf) nodes to subtriangles
- Point location search by testing all (at most 3) children
- RIC: expected depth  $O(\log n)$