Linear Algebra and **Cubes**

Bernd Gärtner **ETH Zürich**

Cube Optimization (I)

Given an *acyclic unique sink orientation*:

acyclic orientation of the n -cube graph, such that every nonempty face has a unique sink.

Orientation might be defined. . .

Cube Optimization (II)

Wanted:

Algorithm for finding the global sink as quickly as possible

We consider simplex-type methods (walks along directed paths); efficiency depends on *pivot* rule being used.

exponentially many steps

polynomially many steps

Linear Algebra. . .

. . . over the reals

is used in constructing *geometric* cube orientations and analyzing *geometric* pivot rules

. . . over finite fields

is used in constructing abstract cube orientations and analyzing combinatorial pivot rules

- Bland's rule
- Random-Edge
- Random-Facet

 \bullet \cdots

Part I: Linear Algebra Over the Reals

Klee & Minty's worst-case linear program, showing that Dantzig's rule may require exponentially many steps

Poor Man's Worst-Case LP

maximize x_n subject to

> $0 \leq x_1 \leq 1$ $\varepsilon x_{i-1} \leq x_i \leq 1 - \varepsilon x_{i-1}, \quad i = 2, \ldots, n$

 $\varepsilon = 1/3$

Observation: Simplex algorithm with "stupid" pivot rule may take $2^n - 1$ steps!

The Klee-Minty Cube (I)

Goal: Fool "smart" pivot rule!

Poor man's LP maximize x_n subject to $0 \leq x_1 \leq 1$ $\varepsilon x_{i-1} \leq x_i \leq 1 - \varepsilon x_{i-1}, \quad i = 2, \ldots, n$

$$
y_1 := x_1
$$

\n
$$
y_i := x_i - \varepsilon x_{i-1}, \quad i = 2, \dots, n
$$

\nslack variables $s_i, \quad i = 1, \dots, n$
\n
$$
\downarrow \qquad \qquad \downarrow
$$

Poor man's LP in standard equality form maximize $\sum_{i=1}^n \varepsilon^{n-i} y_i$ subject to $y_i + 2\sum_{j=1}^{i-1}\varepsilon^{i-j}y_j + s_i = 1, i = 1,\ldots,n$ $y_i \geq 0, \quad i = 1, \ldots, n$ $s_i \geq 0, \quad i = 1, \ldots, n$

The Klee-Minty Cube (II)

The Klee-Minty Cube (III)

Lemma (fooling the stupid): The "stupid" pivot rule of always choosing the variable with smallest positive coefficient in the z-row leads to $2^n - 1$ pivot steps on the poor man's worstcase LP...

. . . while one step suffices for the "smart" pivot rule of always choosing the variable with *largest* positive coefficient.

Lemma (fooling the smart): For $\varepsilon > 3$, the Klee-Minty cube is the tweaked poor man's LP

maximize
$$
\sum_{i=1}^{n} \varepsilon^{n-i} y_i
$$
 subject to
\n
$$
y_i + 2\sum_{j=1}^{i-1} \varepsilon^{i-j} y_j + s_i = \varepsilon^{2i}, i = 1, ..., n
$$
\n
$$
y_i \geq 0, \quad i = 1, ..., n
$$
\n
$$
s_i \geq 0, \quad i = 1, ..., n
$$

It is a cube on which the "smart" pivot rule of always choosing the variable with *largest* positive coefficient in the z-row leads to $2^n - 1$ pivot steps. \leftarrow Dantzig's rule!

Part II: Linear Algebra Over $GF(2)$

- Ignorant (combinatorial) pivot rules
- Combinatorics of the Klee-Minty cube
- Random-Edge on the Klee-Minty cube
- Matoušek's abstract cubes
- Random-Facet on Matoušek cubes

Ignorant Pivot Rules (I)

Random-Edge: among the variables with positive coefficient in the z -row, choose one uniformly at random.

- Behavior only depends on sign pattern in the z-row, not on actual coefficients
- Expected number of steps is the same for the poor man's worst case LP and the Klee-Minty cube

Questions:

- What is this expected number of steps?
- Does the Klee-Minty cube fool the ignorant?

The Combinatorial z -row (I)

$$
\begin{pmatrix} x_n \\ s_n \end{pmatrix} \rightarrow \begin{pmatrix} 1 \\ 0 \end{pmatrix}
$$

$$
\bullet \begin{pmatrix} x_i \\ s_i \end{pmatrix} \rightarrow \begin{pmatrix} 0 \\ 1 \end{pmatrix}, \quad i = 1, \dots, n-1
$$

$$
\bullet \begin{pmatrix} + \\ - \end{pmatrix} \rightarrow \begin{pmatrix} 1 \\ 0 \end{pmatrix}
$$

z-row of optimal tableau:

$$
\begin{array}{c|cc}\n\hline\nz & = & 1 - s_3 - \varepsilon y_2 - \varepsilon^2 y_1 \\
\hline\n\text{vertex} & 0 & 0 & 0 \\
\text{value} & 0 & 0 & 0\n\end{array}
$$

The Combinatorial z -row (II)

Combinatorial KM-cube (I)

- vertices $\equiv GF(2)^n$, values $\equiv GF(2)^n$
- adjacent vertices differ in exactly one coordinate
- vertex v has value Av , where

$$
A = \begin{pmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \\ 1 & 1 & 1 \end{pmatrix} \in GF(2)^{n \times n}.
$$

- \bullet v has better objective function value than v^\prime iff $Av < Av^\prime$ under lexicographic ordering of the values
- the step from v to $v + e_i$ is a legal (improving) pivot iff $(Av)_i = 1$, equivalently if there is an odd number of 1-entries in $v_1 \ldots v_i$.

Combinatorial KM-cube (II)

Random-Edge Revisited

Fast Game (on values): among all 1-entries, choose one uniformly at random and flip it, along with all entries further to the right.

 \Rightarrow expected number of steps is $O(n^2)$

Lower bound???

The Slow Game

Slow Game: among all entries, choose one uniformly at random; if it is a 1-entry, flip it, along with all entries further to the right, otherwise, do nothing.

 \Rightarrow expected number of steps is $\Omega(n^2/\log n)$

Fast vs. Slow Game (I)

• flip of w at i (real or void): $w \rightarrow A^{(i)}w$, where

$$
A^{(i)} = \begin{pmatrix} 1 & & & \downarrow \text{column } i \\ & \ddots & & \\ & & 1 & \\ & & & 0 \\ & & & 1 & 1 \\ & & & & \vdots \\ & & & & 1 \end{pmatrix}
$$

• flip sequence:
$$
s = (i_1, i_2, \ldots)
$$

- \bullet flip of w with $(s,k){\colon \ } w \to w^{(s,k)},$ where $w^{(s,k)} \;\; := \;\; A^{(s,k)} w,$ $A^{(s,k)} = A^{i_k} \cdots A^{i_2} A^{i_1}$
- $S =$ set of flip sequences, $V = GF(2)^n$

Fast vs. Slow Game (II)

• $F(w), S(w)$: expected number of steps in fast and slow game, starting with w

$$
\begin{aligned}\n\bullet \ F(n) &:= \frac{1}{2^n} \sum_{w \in \mathcal{V}} F(w) \\
F(n) &= \sum_{k=1}^{\infty} \text{prob}(w^{(s,k)} \neq w^{(s,k-1)}) \\
&= \sum_{k=1}^{\infty} \frac{1}{n} \sum_{i=1}^n \text{prob}(w^{(s,k-1)}_i = 1) \\
&= \frac{1}{n} \sum_{i=1}^n \sum_{k=1}^{\infty} \text{prob}(A^{(s,k-1)}w)_i = 1) \\
&= \frac{1}{n} \sum_{i=1}^n \sum_{k=1}^{\infty} \text{prob}(e_i^T A^{(s,k-1)})w = 1) \\
&= \frac{1}{2n} \sum_{i=1}^n \sum_{k=1}^{\infty} \text{prob}(e_i^T A^{(s,k-1)} \neq 0) \\
&= \frac{1}{2n} \sum_{i=1}^n \sum_{k=1}^{\infty} S(Ae_{n-i+1})\n\end{aligned}
$$

Random-Edge Performance

Theorem: There is a starting vertex of the n -dimensional Klee-Minty cube for which the simplex algorithm with the Random-Edge pivot rule requires an expected number of

 $\Omega(n^2/\log n)$

steps.

⇒ Klee-Minty cube "mildly" fools Random-Edge

Ignorant Pivot Rules (II)

Random-Facet: among the variables with positive coefficient in the z -row, choose the one whose index comes first in an initially chosen random permutation π of the indices.

- $\pi = (1, 2, \ldots, n)$: "stupid" rule, $2^n - 1$ steps
- $\pi = (n, n-1, ..., 1)$: "smart" rule, 1 step

Theorem: For every starting vertex of the n dimensional Klee-Minty cube, the simplex algorithm with the Random-Facet pivot rule requires $O(n^2)$ steps, and this bound is tight.

Beyond Klee-Minty Cubes

• Klee-Minty cube:

vertex
$$
v \rightarrow
$$
 value $w = \begin{pmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \\ 1 & 1 & 1 \end{pmatrix} v$.

• Matoušek cube:

vertex $v \rightarrow$ value $w = Av$,

with $A \in GF(2)^{n \times n}$ being a lower-triangular, invertible matrix

Questions:

- Are Matoušek cubes combinatorial models of linear programs over (deformed) cubes?
- Does some Matoušek-cube fool Random-Facet? (It won't fool Random-Edge!)

Matoušek Cubes

Examples:

• The Matoušek cube $A = I_n$ is generated by the "unit cube" LP

maximize $\sum_{i=1}^n i \cdot x_i$ subject to

$$
0 \leq x_i \leq 1, \quad i=1,\ldots,n
$$

• The 3-dimensional Matoušek cubes

$$
A_1 = \left(\begin{array}{cc} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 1 & 1 & 1 \end{array}\right), \quad A_2 = \left(\begin{array}{cc} 1 & 0 & 0 \\ 1 & 1 & 0 \\ 0 & 1 & 1 \end{array}\right)
$$

do not come from any LP — they are generalized LPs

Matoušek Cubes — Results

Theorem: For a random starting vertex of a random n -dimensional Matoušek cube, the simplex algorithm with the Random-Facet pivot rule requires an expected number of $e^{\Omega(\sqrt{n})}$ steps, and this bound is tight.

⇒ Matoušek cubes fool Random-Facet

Theorem: For every starting vertex of every LP -induced n-dimensional Matoušek cube, the simplex algorithm with the Random-Facet pivot rule requires $O(n^2)$ steps, and this bound is tight.

⇒ still no LP known to fool Random-Facet

LP-induced Matoušek Cubes

Observation: If the n -dimensional Matoušek cube A is LP-induced, then A does not contain the "forbidden minors"

$$
A_1 = \left(\begin{array}{cc} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 1 & 1 & 1 \end{array}\right), \quad A_2 = \left(\begin{array}{cc} 1 & 0 & 0 \\ 1 & 1 & 0 \\ 0 & 1 & 1 \end{array}\right).
$$

Lemma: Let $A \in GF(2)^{n \times n}$ be a matrix without forbidden minors. Then A^{-1} has at most one off-diagonal one-entry per row.

Corollary: Among the $2^{\binom{n}{2}}$ $_{2}^{n})$ Matoušek cubes, at most $n! \approx 2^{n \log n}$ are LP-induced.

Examples:

Unit cube:

$$
A = \begin{pmatrix} 1 & & & & \\ 0 & 1 & & & \\ 0 & 0 & 1 & & \\ 0 & 0 & 0 & 1 & \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix}, \quad A^{-1} = \begin{pmatrix} 1 & & & & \\ 0 & 1 & & & \\ 0 & 0 & 1 & & \\ 0 & 0 & 0 & 1 & \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix}
$$

Klee-Minty cube:

$$
A = \begin{pmatrix} 1 & 1 & & & \\ 1 & 1 & 1 & & \\ 1 & 1 & 1 & 1 & \\ 1 & 1 & 1 & 1 & 1 \end{pmatrix}, \quad A^{-1} = \begin{pmatrix} 1 & & & & \\ 1 & 1 & & & \\ 0 & 1 & 1 & & \\ 0 & 0 & 1 & 1 & \\ 0 & 0 & 0 & 1 & 1 \end{pmatrix}
$$

Random-Facet: the LP-Case

- Random permutation π of the coordinates, start vertex v
- \bullet Among the coordinates i which are flippable $((Av)_i = 1)$, flip the first one in π ; repeat until 0 is reached
- k: last coordinate in π

 v' : first vertex where no coordinate different from k is flippable

 $v^{\prime\prime}$: successor of v^{\prime} (if existing)

Cheap Case:
$$
v_k = 0
$$

\n $\frac{v}{\pi} \begin{bmatrix} 1 & 0 & 0 & 1 & 1 \\ 1 & 5 & 3 & 4 & 2 \end{bmatrix}$ ⇒ $v' = 0$ 0 0 0 0

\n⇒ done!

Express the given values:
$$
v_k = 1
$$

\n
$$
\frac{v}{\pi} \begin{vmatrix} 1 & 0 & 0 & 1 & 1 \\ 2 & 1 & 5 & 3 & 4 \end{vmatrix} \Rightarrow Av' = 0 \quad 0 \quad 0 \quad 1 \quad 0
$$

•
$$
Av'' = Av' + A_k = e_k + A_k
$$

$$
\bullet \ v'' = A_k^{-1} + \mathbf{e}_k
$$

- \bullet the possible v'' over all k with $v_k=1$ are columns of $A^{-1} - I_n$
- in the LP-induced case, for any $j, v''_j = 1$ for \boldsymbol{a} t most one of the possible v''
- with j the second-to-last coordinate in π , $v^{\prime\prime}$ leads to the cheap case $v^{\prime\prime}_j=0$ with high probability (over the random choice of k)

Beyond $GF(2)$?

- $\mathcal{V} = GF(q)^n$ can be interpreted as the vertex set of a polytope which is the product of n simplices of dimension $q-1$ each
- Any lower-triangular, invertible matrix $A \in$ $GF(q)^{n \times n}$ induces an acyclic unique sink orientation on this polytope
- . . . but relation between fast and slow game does not hold — flips are not linear functions anymore
- Performance of Random-Facet on such orientations? Lower bound of

e $\Omega(\sqrt{n\log(nq)})$

might hold for large q and would be significant