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 $\Omega(n^d)$ Pivots**

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The Criss-Cross Method Can Take $\Omega(n^d)$ Pivots

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Abstract

Using deformed products of arrangements, we construct a family of linear programs with n inequalities in \mathfrak{R}^d on which, in the worst-case, the least-index criss-cross method requires $\Omega(n^d)$ (for fixed d) pivots to reach optimality.

Résumé

En utilisant les produits déformés des polytopes, nous construisons une famille de programmes linéaires avec n contraintes en \mathfrak{R}^d sur lesquels la méthode entrecroisée du plus-petit-indice prend $\Omega(n^d)$ (pour d fixé) pivots pour atteindre le sommet optimal.

Communicated by Professor David Avis, GERAD and School of Computer Science, McGill University.

1 Introduction

What is the maximal number of pivots, $C(d, n)$, (along a path) taken by the *least-index criss-cross method* to solve a linear program in dimension d with n inequality constraints? This question was posed by D. Avis in 2002 (private communication). Clearly

$$C(d, n) \leq \binom{n}{d} - 1 \leq n^d \quad \text{for } n \geq d \quad (1)$$

since the maximal number of vertices of an arrangement is $\binom{n}{d}$. In 1978, even before the birth of the least-index criss-cross method, Avis and Chvátal [1] unknowingly proved its exponential worst-case behaviour by exhibiting an example where the number of pivots taken by the simplex method with Bland's rule on a completely degenerate polytope is bounded from below by the $(2d)^{\text{th}}$ fibonacci number which is of the order $(1.618 \dots)^{2d}$.

$$C(d, n) \geq \frac{1}{\sqrt{5}} \left(\left(\frac{1 + \sqrt{5}}{2} \right)^{2d} - \left(\frac{1 - \sqrt{5}}{2} \right)^{2d} \right) - 1 \quad \text{for } n \geq 2d. \quad (2)$$

The result follows from the observation that the least-index criss-cross method and the simplex method with Bland's rule follow the same pivot path on a completely degenerate polytope. In 1990, Roos [13] marginally improved the lower bound by constructing the first example of a nondegenerate polytope where the criss-cross method requires an exponential number of pivots. In his example, the criss-cross method starts at a vertex of the polytope and the inequalities are ordered so that the criss-cross method path mimics the behaviour of the simplex method with Bland's rule and thus always stays on the boundary of the polytope. He showed that

$$C(d, n) \geq 2^{2d} - 1 \quad \text{for } n \geq 2d. \quad (3)$$

Both of these constructions are variants of the *Klee-Minty* examples [10]. To date, all worst-case constructions for the criss-cross method provide a lower bound on $C(d, n)$ which, asymptotically, leaves a significant gap with the upper bound:

$$C(d, n) \text{ is } \Omega(2^{2d}) \text{ and } O(n^d) \text{ for } n \geq 2d \text{ where } d \text{ is fixed.} \quad (4)$$

In fact, it remained unclear whether the criss-cross method could take a path of length longer than the maximal number of vertices, $M(d, n)$, that a d -polytope with n facets can have.

$$C(d, n) \stackrel{?}{\geq} M(d, n) \quad (= \theta(n^{\lfloor \frac{d}{2} \rfloor})) \quad \text{for fixed } d \quad (5)$$

In the present paper we show how to construct a family of examples which not only answer this question affirmatively, but also show that $C(d, n)$ is $\Omega(n^d)$ for fixed d , implying that nearly every vertex of the arrangement can be visited.

Theorem 1 (Main Theorem) *For fixed d , the function $C(d, n)$ grows like a polynomial of degree d in n :*

$$C(d, n) \text{ is } \theta(n^d) \text{ for } n \geq 2d \quad (6)$$

Our construction uses the powerful tool of a *deformed product of arrangements*, an extension of a *deformed product of polytopes* as defined recently by Amenta and Ziegler [2]. The reader familiar with polytopes, linear programming, and deformed products of polytopes can breeze through section 2. Section 3 examines the behaviour of the least-index criss-cross method on a deformed product of polytopes, and by section 4 we are ready to construct a family of worst-case examples.

2 Preliminaries

2.1 Polyhedra

A d -polyhedron P is the intersection of n closed halfspaces (inequalities) in \mathbb{R}^d or equivalently the convex hull of a finite set of points in \mathbb{R}^d [8, p. 31]. We write these representations as

$$\text{(Halfspace-rep.) } P = \{x \in \mathbb{R}^d : a_i^t x \leq b_i \text{ for } 1 \leq i \leq n\} \text{ where } a_i \in \mathbb{R}^d, \text{ and } b_i \in \mathbb{R}, \quad (7)$$

and

$$\text{(Vertex-rep.) } P = \text{conv}\{p_1, \dots, p_m\} \text{ for points } p_i \in \mathbb{R}^d. \quad (8)$$

The *faces* of P are all the subsets of the form $F = \{x \in P : \alpha^t x = \beta\}$ for some $\alpha \in \mathbb{R}^d$, and $\beta \in \mathbb{R}$, where $\alpha^t x \leq \beta$ is a *valid* inequality for P ; meaning that $\alpha^t x \leq \beta$ is satisfied for all $x \in P$. The faces of P are themselves polyhedra and the faces of dimensions 0, 1, $d-2$, $d-1$, and k are called *vertices*, *edges*, *ridges*, *facets*, and *k-faces* of P . A bounded polyhedron is called a *polytope*. The interested reader is referred to the textbooks by Grünbaum [8] and Ziegler [19] for a comprehensive study of convex polytopes and their properties.

Definition 1 (Simple Polytopes) *A d -polytope is simple if every vertex lies on exactly d facets.*

Definition 2 (Combinatorially Equivalent Polytopes) *Two polytopes P and Q are combinatorially equivalent if there is a bijection between their vertices, $\text{vert}(P) = \{p_1, \dots, p_m\}$ and $\text{vert}(Q) = \{q_1, \dots, q_m\}$, such that for any subset $I \subseteq \{1, \dots, m\}$, the convex hull $\text{conv}\{p_i : i \in I\}$ is a face of P if and only if $\text{conv}\{q_i : i \in I\}$ is a face of Q .*

Definition 3 (Normally Equivalent Polytopes) *Two polytopes P and Q are normally equivalent if they are combinatorially equivalent and each facet $\text{conv}\{p_i : i \in I\}$ of P is parallel to the corresponding facet $\text{conv}\{q_i : i \in I\}$ of Q .*

Theorem 2 (Upper Bound Theorem, McMullen [11]) *A d -dimensional polytope with n facets has no more than*

$$\binom{n - \lceil \frac{d}{2} \rceil}{\lfloor \frac{d}{2} \rfloor} + \binom{n - 1 - \lceil \frac{d-1}{2} \rceil}{\lfloor \frac{d-1}{2} \rfloor} \quad (9)$$

vertices, where equality is attained only by the polars of neighborly polytopes (for example, by the polars of cyclic polytopes).

This upper bound is a polynomial in n of degree $\lfloor \frac{d}{2} \rfloor$ in the case of fixed dimension:

$$M(d, n) = \theta(n^{\lfloor \frac{d}{2} \rfloor}) \quad \text{for fixed } d. \quad (10)$$

2.2 Hyperplane Arrangements

A *hyperplane* is a set $h = \{x \in \mathbb{R}^d : \alpha^t x = \beta\}$ for some nonzero $\alpha \in \mathbb{R}^d$, and $\beta \in \mathbb{R}$. A finite set of hyperplanes H in \mathbb{R}^d induces a decomposition of \mathbb{R}^d into connected open cells called an *arrangement* A_H . The 0, 1, 2, $(d-1)$, and k -dimensional cells of A_H are termed vertices, edges, faces, facets, and k -cells. Two vertices of A_H are *adjacent* if they share $d-1$ hyperplanes, in other words they share an edge. Given a hyperplane arrangement, $A_H \subseteq \mathbb{R}^d$, and a linear functional $\varphi : \mathbb{R}^d \rightarrow \mathbb{R}$ such that π_{\min} and π_{\max} are the vertices of A_H that minimize resp. maximize φ with $0 \leq \varphi(\pi_{\min}) \leq \varphi(\pi_{\max}) \leq 1$, then we write $\varphi(\text{vert}(A_H)) \subseteq [0, 1]$. We will assume that the hyperplanes are labeled as $H = \{h_1, h_2, \dots, h_n\}$, and that hyperplanes are *oriented*: each has a positive side and negative side that are given by $\{x \in \mathbb{R}^d : \alpha^t x > \beta\}$ and $\{x \in \mathbb{R}^d : \alpha^t x < \beta\}$. An arrangement of oriented hyperplanes is an example of an *oriented matroid*, and shares all of its features (see [12]). In particular, each cell of the arrangement is represented as a *signed vector* in $\{+, 0, -\}^n$ indicating the position of the cell with respect to the hyperplanes of H .

Proposition 1 *A polytope $P \subseteq \mathbb{R}^d$ induces an arrangement of oriented hyperplanes A_P .*

Proof. P is defined by the intersection of a finite number of halfspaces, and a halfspace is an oriented hyperplane. Thus P induces an arrangement of oriented hyperplanes A_P where the feasible region P is simply the intersection of the nonnegative sides of A_P . ■

Definition 4 (Combinatorially Equivalent Arrangements) *Two arrangements are combinatorially equivalent if the two sets of sign vectors of cells of the arrangements are exactly the same (i.e. the underlying oriented matroids are exactly the same).*

Definition 5 (Normally Equivalent Arrangements) *Two hyperplane arrangements are said to be normally equivalent if they are combinatorially equivalent and the corresponding unit hyperplane normals coincide.*

For an introduction to arrangements we refer to Halperin [9], and for a study on their combinatorial structure see [12] and [17].

2.3 Linear Programming

Linear Programming is the problem of maximizing a linear functional φ over a polyhedron P :

$$\max \varphi(x) : x \in P \subseteq \mathbb{R}^d. \tag{11}$$

For the remainder of our discussion, we will assume that the linear program is *feasible* (P is non-empty), P is simple, and that φ is bounded by P .

Starting at a vertex defined by the set H of d intersecting hyperplanes, a *pivot* is the operation of exchanging a hyperplane $h_i \in H$ for a hyperplane $h_j \notin H$ that intersects the edge $H \setminus \{h_i\}$. This results in a second vertex defined by $H' := H \setminus \{h_i\} \cup \{h_j\}$ (henceforth abbreviated by $H - h_i + h_j$). *Pivot methods* attempt to solve a linear program, (P, φ) , by moving along the edges of A_P from vertex to adjacent vertex, and *pivot rules* determine which adjacent vertex to travel to. The sequence of vertices visited is called the *pivot path*, and when a pivot rule guarantees convergence to optimality then we say it is *finite*. *Simplex (pivot) methods* try to solve a linear program by pivoting along the boundary edges of P from vertex $v_i \in P$ to vertex $v_j \in P$ such that $\varphi(v_i) < \varphi(v_j)$. For example, *Bland's least-index rule* [3], is a finite pivot rule for the simplex method where the edge chosen at v_i leaves the hyperplane that is indexed smaller than the other hyperplanes at v_i leaving which would offer to increase φ . Chvátal's classic book [4], provides an expository introduction to linear programming. See Terlaky and Zhang [15] for a survey on pivot rules.

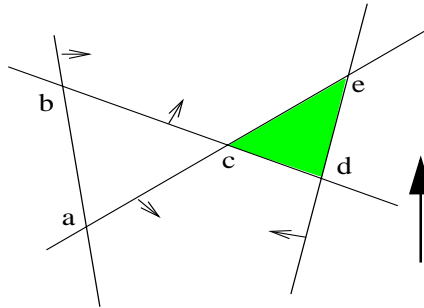


Figure 2.3

Definition 6 (Increasing Edges) For a linear functional $\alpha : \mathbb{R}^d \rightarrow \mathbb{R}$, an edge $[p', p'']$ is α -increasing if $\alpha(p'') > \alpha(p')$.

In Figure 2.3, $[a, b]$, $[a, e]$, and $[d, b]$ are examples of increasing edges.

Definition 7 (Increasing Rays) Given a start point $p \in \mathbb{R}^d$ and a vector $\vec{v} \in \mathbb{R}^d$, a ray $r = (p, \vec{v})$ is the set of all points of the form $p + \lambda \vec{v}$ for all scalar $\lambda \geq 0$. For a linear functional $\alpha : \mathbb{R}^d \rightarrow \mathbb{R}$, a ray is α -increasing if and only if $\alpha(\vec{v}) > 0$.

Definition 8 (Primal Infeasible Vertex) A vertex ξ_i of the arrangement induced by a linear program (P, φ) is primal infeasible if ξ_i is not a vertex of the polytope P . In other words, ξ_i violates at least one inequality of P .

In Figure 2.3, a and b are primal infeasible.

Definition 9 (Dual Infeasible Vertex) A vertex ξ_i , defined by the set H of d intersecting hyperplanes of the arrangement induced by a linear program (P, φ) , is dual infeasible if there is at least one ray starting at ξ_i that is φ -increasing and satisfies all $h \in H$ (every point on the ray lies on the nonnegative side of all hyperplanes $h \in H$).

In Figure 2.3, a, b, c , and d are dual infeasible.

The optimal vertex which maximizes φ is a vertex that is both primal and dual feasible.

2.4 Deformed Products of Polytopes

Recently Amenta and Ziegler [2] presented a construction of deformed products of polytopes that generalized all known constructions of linear programs with “many simplex pivots.”

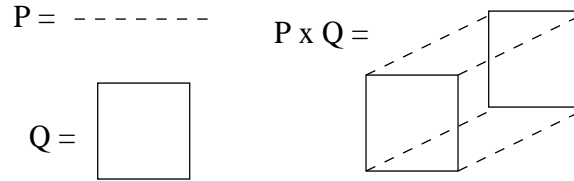
Definition 10 (Product of Polytopes) The product of two polytopes $P \subseteq \mathbb{R}^d$ and $Q \subseteq \mathbb{R}^e$ is given by

$$P \times Q = \left\{ \begin{pmatrix} x \\ u \end{pmatrix} : x \in P, u \in Q \right\}. \tag{12}$$

The vertices of the product are given by

$$\text{vert}(P \times Q) = \left\{ \begin{pmatrix} p_i \\ q_j \end{pmatrix} : p_i \in \text{vert}(P), q_j \in \text{vert}(Q) \right\}, \tag{13}$$

and the facet-defining inequalities for $P \times Q$ are the inequalities of P together with the inequalities of Q . Thus taking the product of two polytopes multiplies the number of vertices and sums the number of facets.



Definition 11 (Deformed Products of Polytopes) Let $P \subseteq \mathbb{R}^d$ be a convex polytope, and $\varphi : P \rightarrow \mathbb{R}$ a linear functional with $\varphi(P) \subseteq [0, 1]$. Let $V, W \subseteq \mathbb{R}^e$ be convex polytopes. Then the deformed product of (P, φ) and of (V, W) is

$$(P, \varphi) \bowtie (V, W) := \left\{ \begin{pmatrix} x \\ v + \varphi(x)(w - v) \end{pmatrix} : \begin{matrix} x \in P \\ v \in V, w \in W \end{matrix} \right\} \subseteq \mathbb{R}^{d+e}. \tag{14}$$

To help visualize the deformed product, it is helpful to observe that when $V = W$, then the deformed product is the standard product:

$$(P, \varphi) \bowtie (V, V) = P \times V, \tag{15}$$

and if we examine a cross-section of the deformed product at some $\varphi(x) = \lambda$, then we will observe the Minkowski sum of λW with $(1 - \lambda)V$.

We now state some facts about deformed products that are proved in [2].

Lemma 1 *Deformed products are convex.*

Theorem 3 *Let $P \subseteq \mathbb{R}^d$ be a d -polytope, $\varphi : \mathbb{R}^d \rightarrow \mathbb{R}$ a linear function with $\varphi(P) \subseteq [0, 1]$, and $V, W \subseteq \mathbb{R}^e$ be normally equivalent e -polytopes, then:*

- If P has m vertices and s facets, and if V and W have n vertices and t facets, then $Q := (P, \varphi) \bowtie (V, W)$ is a $(d + e)$ -polytope with mn vertices and $s + t$ facets.
- For $P = \text{conv}\{p_1, \dots, p_m\}$, $V = \text{conv}\{v_1, \dots, v_n\}$, and $W = \text{conv}\{w_1, \dots, w_n\}$, the deformed product is a polytope given by

$$Q = (P, \varphi) \bowtie (V, W) = \text{conv}\{q(i, j) : 1 \leq i \leq m, 1 \leq j \leq n\}. \quad (16)$$

where the vertices of Q , which we denote $q(i, j)$, are defined as:

$$q(i, j) = \begin{pmatrix} p_i \\ v_j + \varphi(p_i)(w_j - v_j) \end{pmatrix} : \begin{matrix} 1 \leq i \leq m \\ 1 \leq j \leq n. \end{matrix} \quad (17)$$

- If P, V and W are given by

$$\begin{aligned} P &= \{x \in \mathbb{R}^d : a_k x \leq \xi_k \text{ for } 1 \leq k \leq s\}, \\ V &= \{u \in \mathbb{R}^e : b_l u \leq \beta_l \text{ for } 1 \leq l \leq t\}, \text{ and} \\ W &= \{u \in \mathbb{R}^e : b'_l u \leq \beta'_l \text{ for } 1 \leq l \leq t\}, \end{aligned} \quad (18)$$

then the deformed product is given by

$$(P, \varphi) \bowtie (V, W) = \left\{ \begin{pmatrix} x \\ u \end{pmatrix} \in \mathbb{R}^{d+e} : \begin{matrix} a_k x \leq \xi_k \text{ for } 1 \leq k \leq s \\ (\beta_l - \beta'_l)\varphi(x) + b_l u \leq \beta_l \text{ for } 1 \leq l \leq t \end{matrix} \right\}. \quad (19)$$

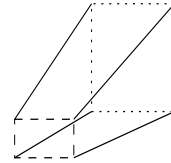
- $Q = (P, \varphi) \bowtie (V, W)$ is combinatorially equivalent to $P \times V$.

$$P = \begin{array}{c} \text{---} \\ | \\ x=0 \quad x=1 \\ | \\ \text{---} \end{array}$$

$$V = \begin{array}{c} \text{---} \\ | \\ \text{---} \\ | \\ \text{---} \end{array}$$

$$W = \begin{array}{c} \text{---} \\ | \\ \text{---} \\ | \\ \text{---} \end{array}$$

$$(P, \varphi) \bowtie (V, W) =$$



Note that Q has two types of edges:

- P -edges of the form $[q(i', j), q(i'', j)]$ for $1 \leq j \leq n$ and for any $1 \leq i', i'' \leq m$ such that $p_{i'}$ and $p_{i''}$ are adjacent vertices of P , and
- (V, W) -edges of the form $[q(i, j'), q(i, j'')]$ for $1 \leq i \leq m$ and for any $1 \leq j', j'' \leq n$ such that $v_{j'}$ and $v_{j''}$ are adjacent vertices of V (or, equivalently, $w_{j'}$ and $w_{j''}$ are adjacent vertices of W).

Definition 12 (Deformed Product Programs) *Let*

$$\max \varphi(x) : x \in P \quad (20)$$

be a linear program in \mathbb{R}^d and let

$$\max \alpha(u) : u \in V \quad \text{and} \quad \max \alpha(u) : u \in W \quad (21)$$

be two normally equivalent linear programs. Define the deformed product program as

$$\max \widehat{\alpha} \begin{pmatrix} x \\ u \end{pmatrix} = \alpha(u) : \begin{pmatrix} x \\ u \end{pmatrix} \in Q = (P, \varphi) \bowtie (V, W). \quad (22)$$

The resulting linear program is the deformed product polytope Q with objective function $\max \alpha(u)$.

Now let's examine the edges of a deformed product program as defined in (22).

Proposition 2 *A P -edge $[q(i', j), q(i'', j)]$ is $\widehat{\alpha}$ -increasing if and only if either $[p', p'']$ is φ -increasing and $\alpha(w_j) > \alpha(v_j)$, or $[p', p'']$ is φ -decreasing and $\alpha(w_j) < \alpha(v_j)$.*

Proposition 3 *A (V, W) -edge $[q(i, j'), q(i, j'')]$ is $\widehat{\alpha}$ -increasing if and only if $[v_{j'}, v_{j''}]$ is α -increasing.*

3 Criss-Cross Methods

Criss-cross methods are pivot methods for solving a linear program (P, φ) whose pivot path can leave the boundary of P . The first criss-cross method was baptized by Zoints [18], and the first finite criss-cross method, the *least-index criss-cross method*, was discovered independently by Terlaky [14], and Wang [16]. The reader is invited to learn about the properties, history, and recent developments pertaining to criss-cross methods by looking at a survey by Fukuda and Terlaky [5].

3.1 The Least-Index Criss-Cross Method

As the name suggests, criss-cross methods have two types of pivots (with respect to an objective function φ): *admissible type I* pivots and *admissible type II* pivots.

Definition 13 (Admissible Type I Pivot) *For every primal infeasible vertex ξ defined by the set H of d intersecting hyperplanes, there exists an oriented hyperplane $h_j \notin H$ that is violated at ξ . A pivot from ξ to vertex ξ' , defined by $H' := H - h_i + h_j$, is an *admissible type I* pivot if ξ' lies on the nonnegative side of h_i .*

If h_j is selected such that j is minimized, followed by selecting h_i to minimize i , then the pivot is a *least-index admissible type I* pivot. In Figure 2.3, a pivot from a to c is of *type I*.

Definition 14 (Admissible Type II Pivot) For every dual infeasible vertex ξ defined by the set H of d intersecting hyperplanes, there exists a φ -increasing ray (ξ, \vec{v}) that lies on an edge $H \setminus \{h_i\}$ that is on the nonnegative side of $h_i \in H$. A pivot from ξ to vertex ξ' , defined by $H' := H - h_i + h_j$, is an admissible type II pivot if there exists a point on (ξ, \vec{v}) that lies on the nonpositive side of h_j .

If h_i is selected such that i is minimized, followed by selecting h_j to minimize j , then the pivot is a *least-index admissible type II* pivot. In Figure 2.3, a pivot from c to e is of type II, as is a pivot from b to a .

We will denote a pivot, exchanging h_i for h_j , by $\text{pivot}(i, j)$. Using these notions, we provide the geometric interpretation of the least-index criss-cross method:

Algorithm 1 (The Least-Index Criss-Cross Method) Given a linear program $(P, \varphi) \subseteq \mathbb{R}^d$, a linear ordering of the inequalities of P , and a vertex ξ of A_P :

Criss-Cross:

If ξ is optimal (both primal feasible and dual feasible) then **Stop**;

If ξ is primal feasible then let $p := +\infty$. Otherwise let $p := j$ such that $\text{pivot}(i, j)$ is the least-index admissible type I pivot from ξ to ξ' . If no pivot exists then the linear program is primal inconsistent, **Stop**;

If ξ is dual feasible then let $q := +\infty$. Otherwise let $q := i'$ such that $\text{pivot}(i', j')$ is the least-index admissible type II pivot from ξ to ξ'' . If no pivot exists then the linear program is dual inconsistent, **Stop**;

If $p < q$, then $\bar{\xi} := \xi'$. Otherwise $\bar{\xi} := \xi''$.

Pivot from ξ to $\bar{\xi}$, let $\xi := \bar{\xi}$ and go to **Criss-Cross**;

Theorem 4 The least-index criss-cross method is finite

Theorem 5 The least-index criss-cross method solves a linear program.

See [7] for simple proofs. From this point forward the criss-cross method will refer to the least-index criss-cross method.

3.2 Deformed Product of Arrangements

Our goal is to construct a family of deformed product programs on which the criss-cross method visits almost all vertices of the arrangement. We begin by analyzing the behaviour of the criss-cross method on the arrangement of hyperplanes of a deformed product program. Hence, we define a *deformed product of arrangements* to be the induced hyperplane arrangement of a deformed product of polytopes, and we extend Theorem 3 to express the properties of the induced hyperplane arrangement of $(Q = (P, \varphi) \bowtie (V, W), \hat{\alpha})$. We facilitate understanding by providing an example (see Figure 3.2) which the reader is encouraged to refer to in order to verify the theorem's statements.

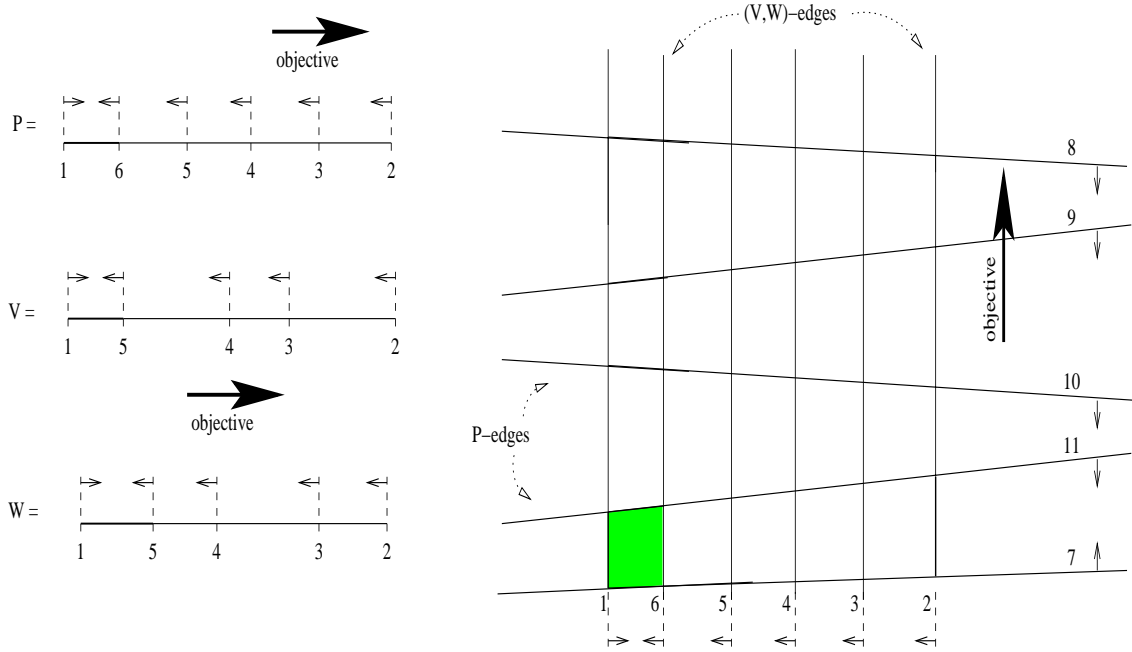


Figure 3.2

Theorem 6 Let $P \subseteq \mathbb{R}^d$ be a d -polyhedron, A_P be the underlying hyperplane arrangement of P , $\varphi : \mathbb{R}^d \rightarrow \mathbb{R}$ a linear function such that $\varphi(\text{vert}(A_P)) \subseteq [0, 1]$, and $V, W \subseteq \mathbb{R}^e$ be normally equivalent e -polyhedra inducing normally equivalent hyperplane arrangements A_V and A_W , then:

- If A_P has \bar{m} vertices and s hyperplanes, and if V and W have \bar{n} vertices and t hyperplanes each, then $Q := (P, \varphi) \boxtimes (V, W)$ is a $(d + e)$ -polytope whose underlying arrangement A_Q has at least $\bar{m} \cdot \bar{n}$ vertices and exactly $s + t$ hyperplanes.
- Specifically, if $\{\pi_1 \dots \pi_m\}$ are vertices of A_P , $\{v_1 \dots v_n\}$ and $\{\omega_1 \dots \omega_n\}$ the vertices of A_V resp. A_W , then we can define $\bar{m} \cdot \bar{n}$ of the vertices of Q , denoted $\gamma(i, j)$, as:

$$\gamma(i, j) = \begin{pmatrix} \pi_i \\ v_j + \varphi(\pi_i)(\omega_j - v_j) \end{pmatrix} : \begin{matrix} 1 \leq i \leq m \\ 1 \leq j \leq n. \end{matrix} \quad (23)$$

- If A_P, A_V , and A_W are given by

$$\begin{aligned} A_P &= \{a_k x \leq \xi_k \text{ for } 1 \leq k \leq s\}, \\ A_V &= \{b_l u \leq \beta_l \text{ for } 1 \leq l \leq t\}, \text{ and} \\ A_W &= \{b'_l u \leq \beta'_l \text{ for } 1 \leq l \leq t\}, \end{aligned} \quad (24)$$

then the arrangement of hyperplanes of the deformed product Q is given by

$$A_Q = \left\{ \begin{matrix} a_k x \leq \xi_k \text{ for } 1 \leq k \leq s \\ (\beta_l - \beta'_l)\varphi(x) + b_l u \leq \beta_l \text{ for } 1 \leq l \leq t \end{matrix} \right\}. \quad (25)$$

- A cell C_Q^i of A_Q is the deformed product of some cell C_P^j with (C_V^k, C_W^k) : $C_Q^i = (C_P^j, \varphi) \boxtimes (C_V^k, C_W^k)$.

Corollary 1 *A cell C_Q^i of the hyperplane arrangement of the deformed product is convex if $0 \leq \varphi(y) \leq 1$ for $y \in C_Q^i$.*

Proof. This follows from Lemma 1 which states that the deformed product of a polytope is convex if $\varphi(P) \subseteq [0, 1]$. ■

Let's examine the edges of a hyperplane arrangement underlying a deformed product program. Note that A_Q has two types of edges:

- A_P -edges of the form $[\gamma(i', j), \gamma(i'', j)]$ for $1 \leq j \leq n$ and for any $1 \leq i', i'' \leq m$ such that $\pi_{i'}$ and $\pi_{i''}$ are adjacent vertices of A_P and
- $A_{(V,W)}$ -edges of the form $[\gamma(i, j'), \gamma(i, j'')]$ for $1 \leq i \leq m$ and for any $1 \leq j', j'' \leq n$ such that $v_{j'}$ and $v_{j''}$ are adjacent vertices of A_V (equivalently, $\omega_{j'}$ and $\omega_{j''}$ are adjacent vertices of A_W).

Proposition 4 *Given a deformed product program (22), an A_P -edge $[\gamma(i', j), \gamma(i'', j)]$ is $\hat{\alpha}$ -increasing if and only if either $[\pi_{i'}, \pi_{i''}]$ is φ -increasing and $\alpha(\omega_j) > \alpha(v_j)$, or $[\pi', \pi'']$ is φ -decreasing and $\alpha(\omega_j) < \alpha(v_j)$.*

Proposition 5 *A (V, W) -edge $[\gamma(i, j'), \gamma(i, j'')]$ is $\hat{\alpha}$ -increasing if and only if $[v_{j'}, v_{j''}]$ is α -increasing.*

The proofs of the preceding statements follow naturally from the proofs given in [2] for deformed products of polytopes. We are now ready to analyze the behaviour of the criss-cross method on deformed product programs.

Corollary 2 (The Criss-Cross Method on Deformed Product Programs) *Let π_1 and π_l be the vertices of P that minimize respectively maximize φ with $0 \leq \varphi(\pi_1) \leq \varphi(\pi_l) \leq 1$. Construct the deformed product program Q , as defined in (22). If we number the inequalities of Q such that the inequalities of P get smaller indices than the inequalities corresponding to (V, W) , then the criss-cross method prefers to pivot along A_P -edges rather than $A_{(V,W)}$ -edges.*

The result is that if the criss-cross method on (P, φ) for the objective function φ takes a path of length l from the π_1 to π_l , and for $-\varphi$ takes a path of length l' from π_l to π_1 , then for $(Q, \hat{\alpha})$ the criss-cross method will follow a path of length l from $\gamma(1, j)$ to $\gamma(l, j)$ if $\alpha(v_j) < \alpha(\omega_j)$, and a path of length l' from $\gamma(l, j)$ to $\gamma(1, j)$ if $\alpha(v_j) > \alpha(\omega_j)$.

4 The Construction

We construct the worst-case example by first building low dimensional examples where the criss-cross method takes many pivots. We then show how to take deformed products of these base cases to construct polyhedra in any dimension where the criss-cross method behaves badly.

Definition 15 Let $C(d, n)$ be the maximal number of pivots taken by the least-index criss-cross method for some linear objective function α on a d -dimensional polyhedron with at most n facets.

Definition 16 Starting at the vertex of P that minimizes α , let $H(d, n)$ be the maximal number of vertices visited along a path taken by the least-index criss-cross method for some linear objective function $\max \alpha$ on the arrangement of hyperplanes induced by a d -dimensional polyhedron with at most n facets.

Clearly $C(d, n) \geq H(d, n) - 1$.

Lemma 2 For $n \geq 2$, $H(1, n) = n$

Proof. Consider the following linear program defined by $\max \varphi = x$ and polytope, given by the inequalities (indexed in order of appearance): $x \geq 0$ and $x \leq (n-i)\lambda$ for $1 \leq i \leq n-1$ and for some constant $\lambda > 0$.

For example, for $n = 6$:

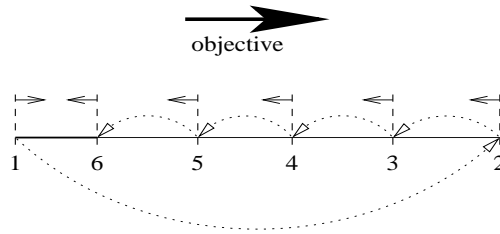


Figure 4a

Let v_i be the vertex defined by hyperplane h_i for $1 \leq i \leq n$. The criss-cross method takes a path of length n from vertex v_1 to vertex v_n when $\varphi = x$. The criss-cross method takes a path of length one from vertex v_n to vertex v_1 when $\varphi = -x$. ■

Note that the construction has $n - 2$ redundant constraints.

Lemma 3 There exists a pair of normally equivalent 1-polytopes, (V, W) , defined by k inequalities each (hence k vertices), and a linear functional α , such that $\alpha(v_i) > \alpha(w_i)$ if i is even and $\alpha(v_i) < \alpha(w_i)$ if i is odd.

Proof. Construct V as in Lemma 2. To build W , for each hyperplane h_i of V , construct h'_i of W by translating h_i in the positive x -direction by (some suitably small) $\epsilon > 0$ if i is odd and by $-\epsilon$ if i is even. The case when $k = 5$ in is illustrated in Figure 3.2. ■

Note that $\alpha(v_k) < \alpha(w_k)$ when k is odd and $\alpha(v_k) > \alpha(w_k)$ if k is even. The following example illustrates the construction of a deformed product and the path that the criss-cross method takes on the underlying arrangement.

Example 1 (See Figure 4b) Construct P (6 inequalities, variable x_1 , $\lambda = 0.1$) and V (5 inequalities, variable x_2) as in Lemma 2, and W (5 inequalities, variable x_2) as in Lemma 3. Let $Q = (P, x) \bowtie (V, W)$ and order the inequalities of Q so that the inequalities coming from P are indexed smaller than those from (V, W) . Consider the path that the criss-cross method takes on the deformed product program $(Q, \hat{\alpha} = x_2)$.

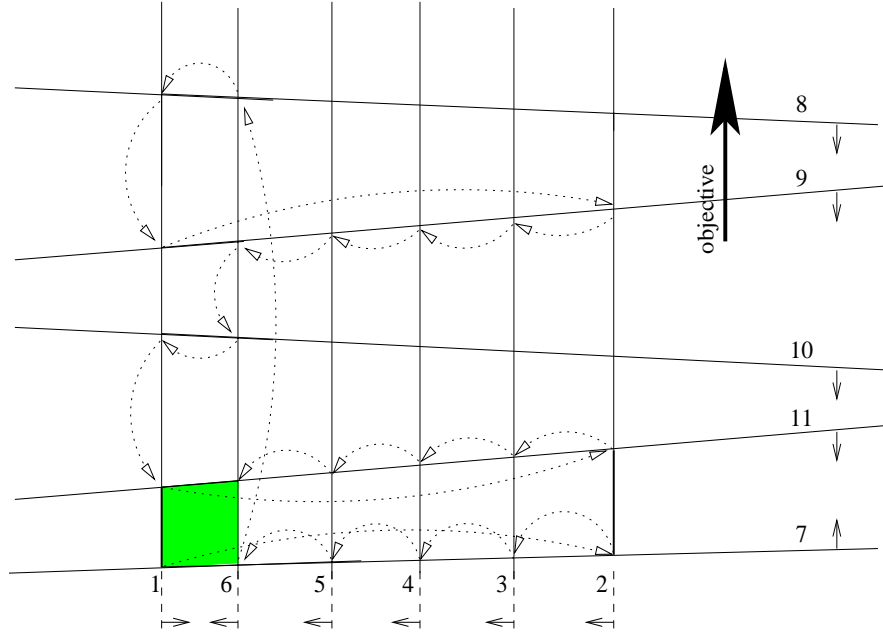


Figure 4b

Theorem 7 For $k \geq 2$ and $n > d \geq 0$,

$$H(d + 1, n + k) \geq \left\lceil \frac{k}{2} \right\rceil \cdot H(d, n) \tag{26}$$

Proof. Take a polytope $P \subseteq \mathbb{R}^d$ with n inequalities for which the least-index criss-cross method for a linear functional $\varphi(x)$ (rescaled such that $\varphi(\text{vert}(A_P)) \subseteq [0, 1]$) follows a criss-cross method path of length $l = H(d, n)$ starting at vertex p_1 and ending at vertex p_l . Now construct the deformed product program

$$\begin{aligned} & \max \alpha \\ & \text{st } \therefore Q = (P, \varphi) \boxtimes (V, W), \end{aligned}$$

where $V, W \subseteq \mathbb{R}$ and α are defined according to Lemma 3. By Corollary 2 we get that the criss-cross method applied to (Q, α) first follows a P -path with l vertices from $\gamma(1, 1)$ to $\gamma(l, 1)$, then after one (V, W) -pivot it follows a P -path of length one from $\gamma(l, 2)$ to $\gamma(1, 2)$, then after one (V, W) -pivot it follows a P -path with l vertices from $\gamma(1, 3)$ to $\gamma(l, 3)$, then after one (V, W) -pivot it follows a P -path of length one from $\gamma(l, 4)$ to $\gamma(1, 4)$, etc. The complete path will thus visit $\lceil \frac{k}{2} \rceil l + \lfloor \frac{k}{2} \rfloor$ vertices arriving at $\gamma(1, k)$ or $\gamma(l, k)$, depending on whether k is even or odd. ■

Remark 1 We could use this result to construct examples, by induction, where $C(d, n)$ is $\Omega(n^d)$ asymptotically for fixed d . However we choose to postpone this analysis since iterative deformed products with the 1-dimensional construction would contain a large number of redundant constraints, in fact $n - 2d$ of them.

Lemma 4 For $n \geq 3$, $H(2, n)$ is $\Omega(n^2)$.

Proof. Consider the following construction: let the i^{th} inequality of P be defined as

$$-2(i - 1)x_1 - (2(n - i) - 1)x_2 \leq -2(i - 1)(2(n - i) - 1). \tag{27}$$

This construction ensures that the x_1 intercept of the i^{th} inequality is greater than the x_1 intercept of the $(i - 1)^{th}$ while the x_2 intercept of the i^{th} inequality is less than that of the $(i - 1)^{th}$ (see Figure 4c).

The least-index criss-cross method on the linear program

$$\begin{aligned} &\text{Maximize } -x_2 \\ &\text{s.t.: } \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \in P, \end{aligned} \tag{28}$$

starting at the vertex defined by the intersection of hyperplanes 1 and n (which we denote $(1, n)$), will take $n - 1$ *type I* pivots $(1, n) \rightarrow (2, n) \rightarrow \dots \rightarrow (n - 1, n)$, and then from $(n - 1, n)$ take one *type II* pivot to $(1, n - 1)$, and then take $n - 2$ *type I* pivots to $(n - 2, n - 1)$, and then one *type II* pivot to $(1, n - 2)$, and then take $n - 3$ *type I* pivots to $(1, n - 3)$, etc. and ending with one *type II* pivot from $(2, 3)$ to $(1, 2)$ visiting a total of $\frac{n(n-1)}{2} = \binom{n}{2}$ vertices. ■

For example, when $n = 7$:

- Maximize $-x_2$
- 1: $0x_1 - 11x_2 \leq 0$
- 2: $-2x_1 - 9x_2 \leq -18$
- 3: $-4x_1 - 7x_2 \leq -28$
- 4: $-6x_1 - 5x_2 \leq -30$
- 5: $-8x_1 - 3x_2 \leq -24$
- 6: $-10x_1 - 1x_2 \leq -10$
- 7: $-12x_1 - 0x_2 \leq 0$

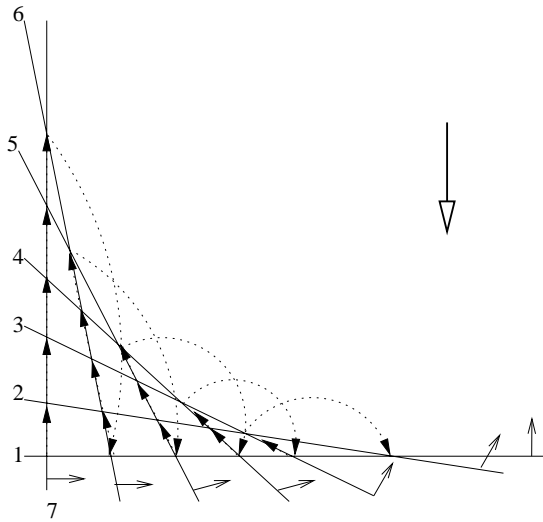


Figure 4c

We offer the following remarks about the construction of Lemma 4:

Remark 2 There are $n - 2$ *type II* pivots, and $\frac{(n-1)(n-2)}{2}$ *type I* pivots.

Remark 3 For every *type I* pivot from intersection (i, j) to (g, j) , if i is odd then g is even, and if i is even then g is odd. Every *type II* pivot has the form (i, j) to $(1, i)$ where $j = i + 1$.

Lemma 5 *There exist normally equivalent 2-dimensional polyhedra V and W with k facets ($k \geq 4$), and objective function $\min \alpha$ for which the criss-cross method takes $\theta(k^2)$ pivots such that corresponding vertices v of V and ω of W , defined by the intersection of hyperplanes h_i and h_j for $i < j$, have the following property:*

$$\alpha(v) > \alpha(\omega) \text{ when } i \text{ is odd, and } \alpha(v) < \alpha(\omega) \text{ when } i \text{ is even.}$$

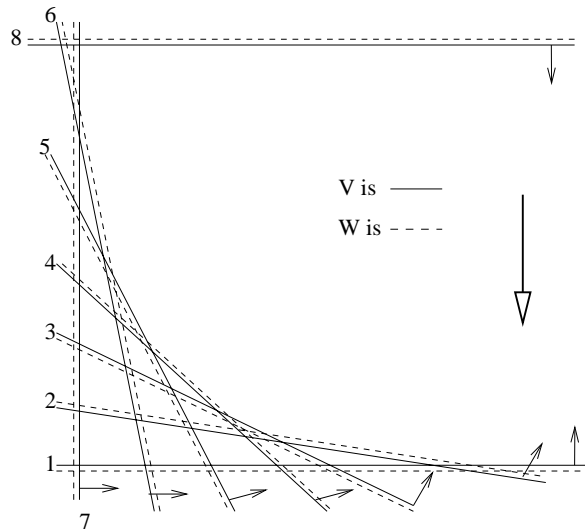


Figure 4d

Proof. Let V be a 2-polyhedron with $k - 1$ inequalities as defined in (27) and define the k^{th} inequality as

$$0x_1 + x_2 \leq 2(k - 2) + C \quad \text{for } C > \frac{k}{2}.$$

This additional inequality ensures that V bounds both α and $-\alpha$. C is chosen such that the x_2 intercept of the k^{th} inequality is greater than that of the $(k - 2)^{th}$. Build W as follows: for $1 \leq i \leq k - 1$ take the i^{th} inequality of V , $b_i x \leq \beta$, and define the i^{th} inequality of W to be $b_i x \leq \beta'$ where $\beta' = \beta - \epsilon$ if i is odd and $\beta' = \beta + \epsilon$ if i is even (see Figure 4d). ϵ is chosen to be positive and suitably small. Let the k^{th} inequality of W be $b_k x \leq \beta'$ where $\beta' = \beta + \epsilon$. Now let's examine corresponding vertices of V and W , v and ω , defined by the intersection of hyperplanes h_i and h_j for $i < j$:

Case 1: i is odd and j odd. This case is illustrated in Figure 4e. n_i and n_j represent the normals of h_i and h_j respectively, or if you wish the direction of translation by ϵ : $|n_i| = |n_j| = \epsilon$. Let $\delta = |d|$, $\theta_1 = \text{angle}(A)$, and $\theta_2 = \text{angle}(B)$. By construction, $0^\circ \leq \theta_1 < \theta_2 \leq 90^\circ$, and $\delta = \epsilon \sin \theta_1$ where $\sin \theta_1 \geq 0^\circ$. Now $\alpha(\omega) < \alpha(v - \delta) \leq \alpha(v)$ which implies $\alpha(v) > \alpha(\omega)$.

Case 2: i is even and j even. This case is symmetric to *case 1*, hence $\alpha(v) < \alpha(\omega)$.

Case 3: i is odd and j even. This case is illustrated in Figure 4f. n_i and n_j represent the normals of h_i and h_j respectively, the direction of translation by ϵ : $|n_i| = |n_j| = \epsilon$. Let $\delta = |d|$, $\theta_1 = \text{angle}(A)$, and $\theta_2 = \text{angle}(B)$. By construction, $0^\circ \leq \theta_1 < \theta_2 \leq 90^\circ$, and $\delta = \epsilon \sin(90^\circ - \theta_2)$ where $\sin(90^\circ - \theta_2) > 0$. Now $\alpha(w) \leq \alpha(v - \delta) < \alpha(v)$ which implies $\alpha(v) > \alpha(\omega)$.

Case 4: i is even and j odd. This case is symmetric to *case 3*, hence $\alpha(v) < \alpha(\omega)$.

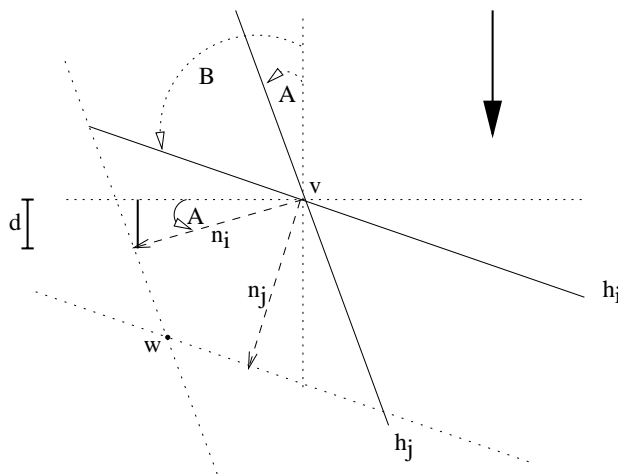


Figure 4e

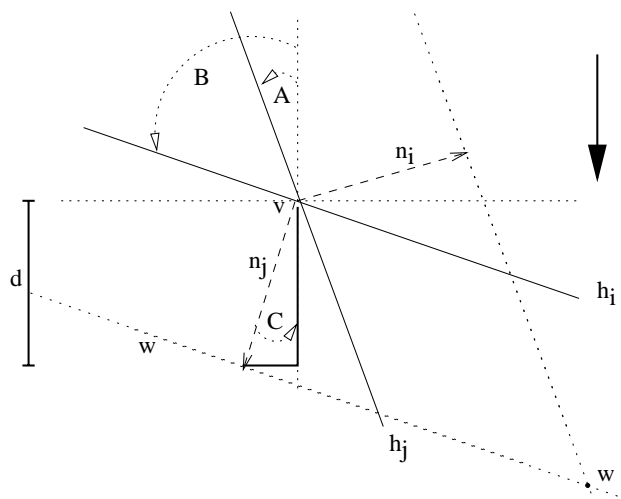


Figure 4f

■

We offer the following remarks about the construction of Lemma 5:

Remark 4 Starting at $(n - 2, n)$ the criss-cross method on V (or W) will take one *type II* pivot to $(1, n)$ and then follow the path described in Lemma 4.

Remark 5 There are $k - 2$ *type II* pivots, and $\frac{(k-2)(k-3)}{2}$ *type I* pivots (see Remark 2 setting $n = k - 1$ and adding one additional *type II* pivot from $(k - 2, k)$ to $(1, k)$).

Definition 17 (Switch Pivot) Given two normally equivalent polyhedra V and W , and a linear objective function α we define a switch pivot to be a pivot from v_i to v_j (w_i to w_j) such that if $\alpha(v_i) > \alpha(w_i)$ then $\alpha(v_j) < \alpha(w_j)$, otherwise if $\alpha(v_i) < \alpha(w_i)$ then $\alpha(v_j) > \alpha(w_j)$.

Lemma 6 Let (V, W) be as defined in Lemma 5. Starting at the intersection of h_{k-1} and h_k the least-index criss cross method takes a $\theta(k^2)$ path to the intersection of h_1 and h_2 on which there are $\theta(k^2)$ switch pivots.

Proof. The least-index criss-cross method on (V, W) takes a $\theta(k^2)$ length path (see Lemma 4). There are five types of pivots with respect to indices of the intersecting hyperplanes of the first vertex v_1 defined by h_i and h_j , and the second vertex v_2 defined by $h_{i'}$ and $h_{j'}$ ($j' = j$ for *type I* pivots):

Pivot	from (i, j) where $i < j$	to (i', j') where $i' < j'$	Switch Pivot?
<i>Type I</i>	i is odd	\rightarrow i' is even	yes
	i is even	\rightarrow i' is odd	yes
<i>Type II</i>	i is odd	\rightarrow $i' = 1$ is odd	no
	i is even	\rightarrow $i' = 1$ is odd	yes
	$(k - 2, k)$	\rightarrow $(1, k)$	yes only if k is even

Thus every *type I* pivot is a switch pivot, and every second *type II* pivot is a switch pivot,

$$\# \text{ of switch pivots} = \begin{cases} \frac{(k-2)(k-3)}{2} + \frac{k-3}{2} & \text{if } k \text{ is odd,} \\ \frac{(k-2)(k-3)}{2} + \frac{k-4}{2} + 1 & \text{if } k \text{ is even,} \end{cases} \quad (29)$$

$$\geq \frac{k^2}{C} \quad \text{for some constant } C > 1 \text{ and all } k \geq 3. \quad (30)$$

The number of switch pivots is $\theta(k^2)$. ■

Theorem 8 For $k \geq 3$, $n > d \geq 0$, and some constant $C > 1$,

$$H(d + 2, n + k) \geq \frac{k^2}{2C} \cdot H(d, n). \quad (31)$$

Proof. Take a polytope $P \subseteq \Re^d$ with n inequalities for which the least-index criss-cross method for a functional $\varphi(x)$ (rescaled such that $\varphi(\text{vert}(A_P)) \subseteq [0, 1]$) follows a criss-cross method path of length $l = H(d, n)$ starting at vertex p_1 and ending at vertex p_l . Let l' be the length of the criss-cross method path from p_l to p_1 for $-\varphi$. Now construct the deformed product program

$$\begin{aligned} & \max \alpha & (32) \\ \text{s.t.: } & Q = (P, \varphi) \boxtimes (V, W), \end{aligned}$$

where V, W and α are defined according to Lemma 5. Let v_{opt} be the optimal vertex of (V, W) . By Corollary 2, we get that the criss-cross method applied to (Q, α) first follows a P -path with l vertices from $\gamma(1, 1)$ to $\gamma(l, 1)$, and then after a (V, W) -switch pivot it follows a P -path of length l' , and then after a (V, W) -switch pivot it follows a P -path of length l , and then after a (V, W) -switch pivot it follows a P -path of length l' , etc. The complete path will visit at least $\frac{1}{2} \frac{k^2}{C} l + \frac{1}{2} \frac{k^2}{C} l'$ vertices ending at (l, opt) . ■

Corollary 3 For $n \geq 2d \geq 2$ and some constant $C > 1$, $C(d, n) = \Omega((\frac{n}{d})^d)$. More specifically:

$$C(d, n) \geq \left\lfloor \frac{2n}{d\sqrt{2C}} \right\rfloor^d \quad \text{if } d \text{ is even,} \quad (33)$$

and

$$C(d, n) \geq \left\lfloor \frac{2n}{(d+1)\sqrt{2C}} \right\rfloor^d \quad \text{if } d \text{ is odd.} \quad (34)$$

Proof. (By induction) Lets begin with the even case, when $d = 2m$ for all $m \geq 0$, let $n = km$:

$$\begin{aligned} H(2m, km) & \geq \frac{k^2}{2C} H(2(m-1), k(m-1)) \\ & \vdots \quad m \text{ times} \\ & \geq \left(\frac{k^2}{2C} \right)^{m-1} H(2, k) \\ & = \frac{k^{2m}}{(2C)^m} \quad \text{by Lemma 4.} \end{aligned}$$

Substituting for $m = \frac{d}{2}$ and $k = \lfloor \frac{2n}{d} \rfloor$, we get

$$H(d, n) \geq \left\lfloor \frac{2n}{d\sqrt{2C}} \right\rfloor^d. \quad (35)$$

For the odd case, when $d = 2m + 1$ for all $m \geq 0$, let $n = k(m + 1)$:

$$\begin{aligned} H(2m + 1, k(m + 1)) &\geq \frac{k^2}{2C} H(2(m - 1) + 1, km) \\ &\quad \vdots \quad m \text{ times} \\ &\geq \left(\frac{k^2}{2C}\right)^m H(1, k) \\ &= \frac{k^{2m+1}}{(2C)^m} \quad \text{by Lemma 2.} \end{aligned}$$

Substituting for $m = \frac{d-1}{2}$ and $k = \left\lfloor \frac{2n}{d+1} \right\rfloor$, we get

$$H(d, n) \geq \left\lfloor \frac{2n}{(d+1)\sqrt{2C}} \right\rfloor^d. \quad (36)$$

The condition $n \geq 2d$ guarantees $k \geq 4$. ■

Remark 6 The construction has no redundant constraints when d is even, and $\left\lfloor \frac{2n}{d+1} \right\rfloor - 2$ redundant constraints when d is odd.

Corollary 4 (Main Theorem) *For fixed dimension d , the function $C(d, n)$ grows like a polynomial of degree d in n :*

$$C(d, n) \text{ is } \theta(n^d) \text{ for } n \geq 2d \text{ where } d \text{ is fixed.} \quad (37)$$

5 Conclusion

Using a construction of deformed product programs, we proved that the worst-case path length that the least-index criss-cross method for solving a linear program can take is $\Omega(n^d)$ for a d -polyhedron defined by n halfspaces (when d is fixed). This result provides a tighter lower bound that asymptotically achieves the upperbound, and also shows that the least-index criss-cross method is worse than simplex methods in the worst case. Despite this negative result, criss-cross methods remain perhaps the best hope of finding a strongly polynomial algorithm for linear programming (see [6]).

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