

## ON THE POLYHEDRAL DECISION PROBLEM\*

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**Abstract.** Computational problems sometimes can be cast in the following form: Given a point  $\mathbf{x}$  in  $R^n$ , determine if  $\mathbf{x}$  lies in some fixed polyhedron. In this paper we give a general lower bound to the complexity of such problems, showing that  $\frac{1}{2} \log_2 f_s$  linear comparisons are needed in the worst case, for any polyhedron with  $f_s$   $s$ -dimensional faces. For polyhedra with abundant faces, this leads to lower bounds nonlinear in  $n$ , the number of variables.

**Key words.** adversary strategy, complexity, dimension, edge, face, linear decision tree, lower bound, polyhedron

**1. Introduction.** Computational problems sometimes can be cast in the following form. Given  $n$  numbers  $x_1, x_2, \dots, x_n$ , determine if they satisfy some fixed set of linear inequalities, i.e., if the point  $\mathbf{x} = (x_1, x_2, \dots, x_n)$  lies in some "polyhedron". For example, the problem of verifying a maximum element can be stated as "Given  $x_1, x_2, \dots, x_n$ , determine if  $x_1 \geq x_i$  for all  $i$ ." As another example, a version of the minimum spanning tree verification problem is the following: Given a weight function  $w$  on the set of edges in a graph  $G$ , determine if  $w(T_0) \leq w(T)$  for all spanning trees  $T$  of  $G$  ( $T_0$  is a fixed spanning tree, and  $w(T)$  is the sum of edge weights in  $T$ ). The aim of this paper is to establish a general lower bound on this type of problems, in terms of some intrinsic characteristics of the polyhedron in question. In contrast to a previous result of this type (Rabin [5]), the present bound can give values larger than the number of variables.

**2. Definitions and notations.** Let  $R^n$  be the space of real  $n$ -tuples. A set  $P$  in  $R^n$  is a *polyhedron* if  $P = \{\mathbf{x} | \mathbf{x} \in R^n, l_i(\mathbf{x}) \leq 0, i = 1, 2, \dots, m\}$ , where  $m$  is an integer,  $\mathbf{x} = (x_1, x_2, \dots, x_n)$ , and  $l_i(\mathbf{x}) = \sum_{1 \leq j \leq n} c_{ij}x_j - a_i$  for some real numbers  $c_{ij}, a_i$ . The *polyhedral decision problem*  $B(P)$  is to determine whether  $\mathbf{x} \in P$  for any input  $\mathbf{x}$ . We are interested in the *linear decision tree model* [1], [5], [10]. An algorithm is a ternary tree with each internal node representing a test of the form " $\sum \lambda_i x_i - c : 0$ ", and each leaf containing a "yes" or "no" answer. For any input, the algorithm proceeds by moving down the tree, testing and branching according to the test results ( $<$ ,  $=$ , or  $>$ ), until a leaf is reached. At that point, the answer to the question "Is  $\mathbf{x} \in P$ ?" is supplied by the leaf. The *cost* of an algorithm is the height of the tree, i.e., the maximum number of tests made for any input. The *complexity* of  $B(P)$  is the minimum cost of any algorithm, and is denoted by  $C(P)$ .

**Faces of a polyhedron.** Let  $P = \{\mathbf{x} | l_i(\mathbf{x}) \leq 0, i = 1, 2, \dots, m\}$  be a polyhedron in  $R^n$ . To each subset  $H$  (maybe  $\emptyset$ ) of  $\{1, 2, \dots, m\}$ , we define a set  $F_H(P) \subseteq R^n$  by  $F_H(P) = \{\mathbf{x} | l_i(\mathbf{x}) < 0 \text{ for each } i \in H; l_i(\mathbf{x}) = 0 \text{ for each } i \notin H\}$ . We say that  $F_H(P)$  is a *face* of dimension  $s$  if the smallest affine subspace of  $R^n$  containing  $F_H(P)$  has dimension  $s$ . (An *affine subspace* is the solution to a set of inhomogeneous equations. See, for example, [6] for more discussions.) The empty face has dimension  $-1$  by convention. Let  $\mathcal{F}_s(P)$  be the set of faces of dimension  $s$  of  $P$ . Note that no two elements of  $\mathcal{F}_s(P)$

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overlap. The set of faces  $\mathcal{F}_s(P)$  is independent of the choice of  $l_i(x)$ . That is, if  $P = \{\mathbf{x} | l'_i(\mathbf{x}) \leq 0, i = 1, 2, \dots, m\}$ , the set  $\mathcal{F}_s(P)$  constructed using  $\{l'_i(\mathbf{x})\}$  is the same as the one constructed using  $\{l_i(\mathbf{x})\}$ . For an intrinsic definition of faces, see for example [3], [8]. A face of dimension 1 is called an edge, as it is part of a line (agreeing with intuition).

**Open polyhedra.** A nonempty set  $Q$  in  $R^n$  is called an *open polyhedron* if  $Q = \{x | l_i(x) < 0, i = 1, 2, \dots, m\}$ . The concepts of faces and set of faces are defined identically as for polyhedra. More precisely, let  $P = \{\mathbf{x} | l_i(\mathbf{x}) \leq 0, i = 1, 2, \dots, m\}$ , then  $F_H(Q) = F_H(P), \mathcal{F}_s(Q) = \mathcal{F}_s(P)$ .

**3. Lower bounds for polyhedral decision problems.** Let  $T$  be a polygon on the plane. Suppose we are asked to decide if a given point  $x$  is inside  $T$  by making a series of tests of the form “ $\lambda \cdot \mathbf{x} - c : 0$ ”. It is easy to see that about  $\log v$  tests are necessary if  $T$  has  $v$  vertices. Our main result is the following generalization.

**THEOREM 1.** *Let  $P = \{\mathbf{x} | l_i(\mathbf{x}) \leq 0 \text{ for } i = 1, 2, \dots, m\}$  be a polyhedron in  $R^n$ . Then for each  $s$ ,*

$$2^{C(P)} \cdot \binom{C(P)}{n-s} \geq |\mathcal{F}_s(P)|.$$

**COROLLARY.**

$$C(P) \geq \frac{1}{2} \log_2 |\mathcal{F}_s(P)|.$$

Theorem 1 relates the complexity of  $B(P)$  to certain “static” combinatorial properties of the polyhedron  $P$ . Informally, if a polyhedron  $P$  has many edges (or faces), then the theorem says it is difficult to decide whether a point lies in  $P$ . The rest of this section is devoted to proving Theorem 1. Note that the corollary follows from Theorem 1 since

$$\binom{C(P)}{n-s} \leq 2^{C(P)}.$$

We will assume in what follows that  $P$  is of dimension  $n$ . The following informal argument demonstrates that this can be done without loss of generality. Suppose that  $\dim(P) = n' < n$ . Let  $S \subseteq R^n$  be the smallest affine subspace of  $R^n$  containing all of  $P$ ; thus  $\dim(S) = n'$ . Now every test  $\sum \lambda_i x_i - c : 0$  in  $R^n$  either corresponds to a linear test  $\sum \lambda'_i x'_i - c' : 0$  in  $S$  (where  $\mathbf{x}'$  is, for  $\mathbf{x} \in S$ ,  $\mathbf{x}$  expressed in a basis for  $S$ ), or else (if  $\{\mathbf{x} \in R^n | \sum \lambda_i x_i = c\} \supseteq S$ ) the test  $\sum \lambda_i x_i - c : 0$  is useful only for determining if  $\mathbf{x} \in S$ , and not for telling if  $\mathbf{x} \in P$  under the assumption that  $\mathbf{x} \in S$ . Therefore the complexity of determining if an  $\mathbf{x} \in R^n$  is in  $P$  is at least as great as the complexity of determining if an  $\mathbf{x} \in S$  is in  $P$ . Since  $\dim(S) = \dim(P)$  we are finished with our demonstration.

To prove Theorem 1 we shall adopt the “adversary approach” commonly used in deriving lower bounds for decision trees. We shall design an adversary strategy  $\mathcal{A}$  which, for any algorithm, will specify the outcomes for successive queries based on the results of previous queries. The following lemma is essential to the construction of  $\mathcal{A}$ .

**LEMMA 1.** *Let  $Q = \{\mathbf{x} | p_i(\mathbf{x}) < 0, i = 1, 2, \dots, t\}$  be a nonempty open polyhedron,  $q(\mathbf{x}) = \sum_{i=1}^n \lambda_i x_i - c$  a linear form,  $Q_1 = Q \cap \{\mathbf{x} | q(\mathbf{x}) < 0\}$  and  $Q_2 = Q \cap \{\mathbf{x} | q(\mathbf{x}) > 0\}$ . Then for each  $s$ , there exists a  $j \in \{1, 2\}$  such that  $Q_j$  is nonempty, and  $|\mathcal{F}_s(Q_j)| \geq \frac{1}{2} |\mathcal{F}_s(Q)|$ .*

*Proof of Lemma 1.* If  $Q_2 = \emptyset$ , then  $Q \subseteq \{\mathbf{x} | q(\mathbf{x}) \leq 0\}$ . Since  $Q$  is an open set, we must have  $Q \subseteq \{\mathbf{x} | q(\mathbf{x}) < 0\}$ . Therefore,  $Q_1 = Q$ , and  $j = 1$  satisfies the requirements. Similarly, for the case  $Q_1 = \emptyset$  we can choose  $j = 2$ . It remains to prove the lemma when both  $Q_1$  and  $Q_2$  are nonempty. We shall accomplish this by constructing a 1–1 mapping  $\psi$  from  $\mathcal{F}_s(Q)$  into  $\mathcal{F}_s(Q_1) \cup \mathcal{F}_s(Q_2)$ . This then implies that  $|\mathcal{F}_s(Q)| \leq |\mathcal{F}_s(Q_1)| + |\mathcal{F}_s(Q_2)|$ . We can then choose a  $j$  such that  $|\mathcal{F}_s(Q_j)| \geq \frac{1}{2} |\mathcal{F}_s(Q)|$ .

Now we construct  $\psi$ . Let  $F_H(Q) \in \mathcal{F}_s(Q)$ . Define

$$\begin{aligned} A_1 &= F_H(Q) \cap \{\mathbf{x} | q(\mathbf{x}) < 0\}, \\ A_2 &= F_H(Q) \cap \{\mathbf{x} | q(\mathbf{x}) > 0\}, \\ A_3 &= F_H(Q) \cap \{\mathbf{x} | q(\mathbf{x}) = 0\}. \end{aligned}$$

Case (1).  $A_1 \cup A_2 = \emptyset$ . In this case  $F_H(Q) \subseteq \{\mathbf{x} | q(\mathbf{x}) = 0\}$ . Let us write  $Q_1 = \{\mathbf{x} | p_i(\mathbf{x}) < 0, i = 1, 2, \dots, t+1\}$ , with  $p_{t+1}(\mathbf{x}) = q(\mathbf{x})$ . Clearly  $F_H(Q_1) = F_H(Q) \cap \{q(\mathbf{x}) = 0\} = F_H(Q)$ . Define  $\psi(F_H(Q)) = F_H(Q_1)$ .

Case (2).  $A_1 \cup A_2 \neq \emptyset$ . Assume that  $A_1 \neq \emptyset$ ; the case  $A_2 \neq \emptyset$  can be similarly treated. Write as before,  $Q_1 = \{\mathbf{x} | p_i(\mathbf{x}) < 0, i = 1, 2, \dots, t+1\}$  with  $p_{t+1}(\mathbf{x}) = q(\mathbf{x})$ . Define  $H' = H \cup \{t+1\}$ . Clearly  $F_{H'}(Q_1) = F_H(Q) \cap \{x | q(x) < 0\}$  is nonempty and is an  $s$ -dimensional face of  $Q_1$ .

Define  $\psi(F_H(Q)) = F_{H'}(Q_1)$ .

It remains to show that the  $\psi$  constructed is an 1-1 mapping. It is easily seen that  $\psi(F_H(Q)) \subseteq F_H(Q)$ . Since all the  $F_H(Q)$  in  $\mathcal{F}_s(Q)$  are disjoint, it follows that all the  $\psi(F_H(Q))$  are disjoint, hence distinct. This completes the proof of Lemma 1.  $\square$

It would be interesting to know if the same value of  $j$  can be used for every value of  $s$  in Lemma 1.

**The adversary strategy  $\mathcal{A}$ .** The adversary  $\mathcal{A}$  will specify a way to answer questions with the help of a sequence of open polyhedra  $V_0, V_1, V_2, \dots$ . Initially,  $V_0 = Q$  where  $Q = \{\mathbf{x} | l_i(\mathbf{x}) < 0, i = 1, 2, \dots, m\}$ . That  $Q$  is an open polyhedron (i.e.,  $Q \neq \emptyset$ ) is a consequence of the assumption that  $P$  has dimension  $n$  (see e.g. [8, Lemma (2.3.10)]). When the  $j$ th query “ $q_j(x): 0$ ” is asked,  $\mathcal{A}$  has constructed  $V_0, V_1, \dots, V_{j-1}$ . The adversary  $\mathcal{A}$  will decide the outcome and construct  $V_j$  in the following way: Let  $Q_1 = V_{j-1} \cap \{\mathbf{x} | q_j(\mathbf{x}) < 0\}$ , and  $Q_2 = V_{j-1} \cap \{\mathbf{x} | q_j(\mathbf{x}) > 0\}$ ; by Lemma 1, there is an  $i \in \{1, 2\}$  such that  $Q_i \neq \emptyset$ , and  $|\mathcal{F}_s(Q_i)| \geq \frac{1}{2} |\mathcal{F}_s(V_{j-1})|$ ; the adversary’s answer to the  $j$ th query is then “ $q_j < 0$ ” if  $i = 1$ , and “ $q_j > 0$ ” if  $i = 2$ ;  $V_j$  is defined to be  $Q_i$ .

**Analysis of the adversary strategy.** Let  $q_j(x): 0$  ( $j = 1, 2, \dots, t$ ) be the entire sequence of queries asked by the algorithm faced with outcomes determined by  $\mathcal{A}$ . Let  $\varepsilon_j q_j(\mathbf{x}) < 0$  be the results of the queries ( $\varepsilon_j = \pm 1$ ). Then,

$$(1) \quad V_i = \{\mathbf{x} | l_i(\mathbf{x}) < 0, i = 1, 2, \dots, m, \varepsilon_j q_j(\mathbf{x}) < 0, j = 1, 2, \dots, t\} \neq \emptyset$$

and

$$|\mathcal{F}_s(V_i)| \geq \frac{1}{2} |\mathcal{F}_s(V_{i-1})| \geq \frac{1}{2^2} |\mathcal{F}_s(V_{i-2})| \geq \dots \geq \frac{1}{2^i} |\mathcal{F}_s(V_0)|, \text{ i.e.,}$$

$$(2) \quad |\mathcal{F}_s(V_i)| \geq \frac{1}{2^i} |\mathcal{F}_s(Q)|.$$

For each  $\mathbf{x} \in V_i$ , the same leaf in the tree  $T$  is reached and the algorithm must say “yes,  $\mathbf{x} \in P$ ”. Since the algorithm only knows that  $\mathbf{x} \in \{\mathbf{x} | \varepsilon_j q_j(\mathbf{x}) < 0, j = 1, 2, \dots, t\}$ , we have

$$\{\mathbf{x} | \varepsilon_j q_j(\mathbf{x}) < 0, j = 1, 2, \dots, t\} \subseteq P.$$

As  $Q$  is the “largest” open set contained in  $P$ , we have

$$\begin{aligned} \{\mathbf{x} | \varepsilon_j q_j(\mathbf{x}) < 0, j = 1, 2, \dots, t\} &\subseteq Q \\ &= \{\mathbf{x} | l_i(\mathbf{x}) < 0, i = 1, 2, \dots, m\}. \end{aligned}$$

Therefore, (1) can be written as

$$(3) \quad V_t = \{\mathbf{x} \mid \varepsilon_j q_j(\mathbf{x}) < 0, j = 1, 2, \dots, t\}.$$

As there are only  $t$  linear functions in (3), there can be at most  $\binom{t}{n-s}$   $s$ -dimensional faces of  $V_t$ . Therefore,

$$(4) \quad \binom{t}{n-s} \cong |\mathcal{F}_s(V_t)|.$$

Equations (2) and (4) lead to

$$(5) \quad 2^t \cdot \binom{t}{n-s} \cong |\mathcal{F}_s(V_t)|.$$

As the left-hand side of (5) is an increasing function of  $t$ , and  $C(P) \cong t$ , we have proved Theorem 1.  $\square$

**4. Remarks.** General discussions on the maximum number of faces that a polyhedron can have are given in [3] and [7]. As there can be  $\approx \binom{m}{n-1}$  edges for certain polyhedra defined by  $m$  inequalities, the corollary to Theorem 1 establishes a lower bound of order  $n \log m$  for, say  $m > n^2$ , to the corresponding polyhedral decision problem.

It would be interesting to find a “natural” problem in concrete computational complexity for which the bound of Theorem 1 yields a nontrivial (i.e., nonlinear) lower bound. In this regard we mention that, originally, it was hoped that the present approach would lead to an  $\Omega(n^2 \log n)$  lower bound to the complexity of the all-pair shortest paths problem. That bound would follow if the *triangular polyhedron*  $P^{(n)}$  in  $R^{(2)}$ , defined as  $\{\mathbf{x} \mid \mathbf{x} = (x_{ij} \mid 1 \leq i < j \leq n); x_{ik} \geq 0, x_{ij} + x_{jk} \geq x_{ik}$  for all  $1 \leq i < k \leq n$  and  $1 \leq j \leq n\}$  (we define  $x_{ii} = 0$  and  $x_{ij} = x_{ji}$ , if  $i > j$ ), has at least  $\exp(cn^2 \log n)$  edges<sup>1</sup>. However, it has recently been shown by Graham, Yao, and Yao [2] that  $P^{(n)}$  has less than  $\exp(cn^2)$  edges, with the implication that only a  $cn^2$  lower bound can be obtained in this approach.

One candidate for the application of Theorem 1 is the problem of constructing optimal alphabetic trees [4], for which the best algorithm known has an  $O(n \log n)$  running time. For a start, what is the number of edges in the polyhedron corresponding to deciding if a complete balanced tree is an optimal alphabetic tree? Another candidate is the verification problem for minimum spanning trees mentioned in the Introduction. It seems difficult, however, to obtain a nonlinear bound in this case, since the number of edges involved is no more than  $\exp(cn \log^* n)$  (because the problem can be solved in  $O(n \log^* n)$  by Tarjan’s result [9]).

<sup>1</sup> See [11] for a proof of this statement. We remark that it was incorrectly stated in [11] that  $P^{(n)}$  has provably  $\exp(cn^2 \log n)$  edges.

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