



On the Complexity of the Class of Regions Computable by a Two-Layered Perceptron

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Abstract. This work is concerned with the computational complexity of the recognition of \mathcal{LP}_2 , the class of regions of the Euclidian space that can be classified exactly by a two-layered perceptron. Several subclasses of \mathcal{LP}_2 of particular interest are also considered. We show that the recognition problems of \mathcal{LP}_2 and of other classes considered here are intractable, even in some favorable circumstances. We then identify special cases having polynomial time algorithms.

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1 Introduction

Several works have been lately devoted to the problem of characterizing the regions of the Euclidian space \mathbb{R}^n that can be classified by a multilayer perceptron (MLP) with n real inputs, linear threshold processing units and a single binary output [Lip87, GC90, ZAW91, Zwi94]. Usually, the attention is restricted to \mathcal{U} denoting the set of all regions of \mathbb{R}^n expressible as a finite union of finite intersections of open or closed half-spaces, since the classification of any region out of \mathcal{U} would required infinitely many threshold units.

The construction of a depth-3 MLP (2 hidden layers) classifying any region of \mathcal{U} is straitforward, given any expression of the latter as a union of intersections of half-spaces [Lip87]. Thus, the research in this area is mainly concerned with the characterization of the regions classifiable by a depth-2 MLP. In some works, exact classifications is studied [ZAW92, Sho93, Zwi94], while others deal with almost everywhere correct classifications [Gib96] or approximations [SGM96].

The present work focuses on the computational complexity of the decision problem "Given V, does V belong to \mathcal{LP}_2 ?", where \mathcal{LP}_2 is the class of regions of \mathbb{R}^n exactly classifiable with a depth-2 MLP. The same issue will be studied for some subclasses of \mathcal{LP}_2 and for different input formats for the specification of V and thus, the boundary between tractable and untractable problems will highlight the real core of the difficulty of the general decision problem.

Several constructive training algorithms have been proposed for MLPs. The most elaborate ones (see for example [TTK93]) have at each iteration a strategy of construction based on the causes of the non-membership of a region into the class \mathcal{LP}_2 or some subclasses that will be considered below. The computational complexity of these algorithms is directly related to the complexity of the recognition problems studied in this paper.

2 Formalization of the recognition problems

The functions computed by two-layered perceptrons as considered in the present paper are defined from the n dimensional Euclidian space \mathbb{R}^n onto the Boolean set $\{0,1\}$ and are of the form:

$$f(\boldsymbol{x}) = g(h_1(\boldsymbol{x}), \dots, h_m(\boldsymbol{x})), \qquad (1)$$

$$h_i(\boldsymbol{x}) = \operatorname{sgn}(w_{i0} + \boldsymbol{x}^{\mathsf{T}} \boldsymbol{w}_i), \ i = 1, \dots, m,$$
 (2)

$$q(\mathbf{b}) = \operatorname{sgn}(t_0 + \mathbf{b}^{\mathsf{T}} \mathbf{t}) . \tag{3}$$

In what follows, the coefficients w_{ij} and t_i , $i=1,\ldots,m, j=1,\ldots,n$ and the thresholds w_{i0} and t_0 are real values. The sign function is defined as $\operatorname{sgn}(r)=1$ if $r\geq 0$ and 0 otherwise. Note that replacing this function sgn by a function sgn^d everywhere identical to sgn except in 0 ($\operatorname{sgn}^d(0)=0$) has no incidence on the class \mathcal{LP}_2 , since on the one hand equation $\operatorname{sgn}(r)=-\operatorname{sgn}^d(-r)$ provides a solution in case sgn^d replaces sgn in (2), and on the other hand t_0 can always be chosen distinct from $-\boldsymbol{b}^{\top}\boldsymbol{t}$ for all $\boldsymbol{b}\in\{0,1\}^n$.

Definition 2.1 A region $V \subset \mathbb{R}^n$ is in \mathcal{LP}_2 if and only if its characteristic function can be written in the form of equations (1-3).

A polyhedron (resp. a pseudo-polyhedron) is a finite intersection of closed (resp. open or closed) half-spaces. The set of all half-spaces used in the expression of V as a union of pseudo-polyhedron is called a basis of V. The general recognition problem can be stated as:

 \mathcal{LP}_2 -RECOGNITION: Given $V \in \mathcal{U}$ specified by a basis \mathcal{H} and a union of intersections of elements of \mathcal{H} , is V in \mathcal{LP}_2 ?

There exist simple regions of \mathbb{R}^2 which are in \mathcal{LP}_2 and for which the only MLPs we know of for their classification have a number of hidden units exponential in the size of a reasonable encoding of the region. This observation leads us naturally to a simplified version of \mathcal{LP}_2 -RECOGNITION where each hidden unit of the MLP classifying the region has to compute the characteristic function of one element of the basis given as input of the problem:

 $\overline{\mathcal{LP}}_2$ -RECOGNITION: Given $V \in \mathcal{U}$ specified by a basis \mathcal{H} and a union of intersections of elements of \mathcal{H} , is there an expression of the form of equations (1-3) for the characteristic function of V, such that each h_i is the characteristic function of one element of \mathcal{H} ?

Based on this new recognition problem a sub-class of \mathcal{LP}_2 can be defined, although it requires some care in order to have a basis uniquely specified by the region. A basis \mathcal{H} of V is minimal if V cannot be expressed as a union of pseudo-polyhedra based on a proper subset of \mathcal{H} . A pseudo-polyhedron is fully dimensioned in \mathbb{R}^n if it contains an open ball of dimension n. $V \in \mathcal{U}$ is fully dimensioned if it is a union of fully dimensioned pseudo-polyhedra, and \mathcal{FD} denotes the set of fully dimensioned regions of \mathcal{U} . It can be shown that $V \in \mathcal{FD}$ has a unique minimal basis which is denoted \mathcal{H}_V^{-1} . Restricting our attention to fully dimensioned regions, we can now define the class $\overline{\mathcal{LP}}_2$ as follows:

Definition 2.2 A region V is in $\overline{\mathcal{LP}}_2$ if $V \in \mathcal{FD}$ and if \mathcal{H}_V together with an expression of V as a union of intersections of elements of \mathcal{H}_V is a positive instance of $\overline{\mathcal{LP}}_2$ -RECOGNITION.

In their attempt at characterizing the class \mathcal{LP}_2 , two groups of researchers proposed independently a sufficient condition for a region V to be in \mathcal{LP}_2 [ZAW92, Sho93].

Definition 2.3 A region $V \in \mathbb{R}^n$ is in \mathcal{D} (resp. $\tilde{\mathcal{D}}$) if it can be expressed as an iterated difference of finitely many polyhedra (resp. pseudo-polyhedra) P_1, \ldots, P_k as follows:

$$V = P_1 \backslash (P_2 \backslash (\dots P_k) \dots).$$

Clearly \mathcal{D} is a proper subset of $\tilde{\mathcal{D}}$. The proof that $\tilde{\mathcal{D}} \subset \mathcal{LP}_2$ is established with the following straitforward lemma:

Lemma 1 \forall pseudo-polyhedron P and \forall $V \in \mathcal{LP}_2$, $P \setminus V \in \mathcal{LP}_2$.

Several examples are proposed in the literature showing that $\tilde{\mathcal{D}} \subsetneq \mathcal{LP}_2$. Recently, M. Ramana and the author proved that if $V \in \tilde{\mathcal{D}}$ has a basis \mathcal{H} , there exist polyhedra P_i , i, \ldots, k , all based on \mathcal{H} , such that $V = P_1 \setminus (P_2 \setminus \ldots P_k) \ldots$, which implies that $\tilde{\mathcal{D}} \cap \mathcal{FD} \subset \overline{\mathcal{LP}}_2$.

3 From geometry to combinatorics

An ordered set of half-spaces $\mathcal{H} = \{H_1, \dots, H_h\}$ of \mathbb{R}^n , called an arrangement, defines a mapping $\Phi_{\mathcal{H}} : \mathbb{R}^n \to \{0,1\}^h$, where $(\Phi_{\mathcal{H}})_i$ is the characteristic function of H_i . The set $D_{\mathcal{H}} = \Phi_{\mathcal{H}}(\mathbb{R}^n)$ is called the domain of the arrangement \mathcal{H} . Given an arrangement \mathcal{H} , there is a one-to-one correspondence between regions of \mathcal{U} having \mathcal{H} as basis and partial Boolean functions defined on $D_{\mathcal{H}}$: the function corresponding $V \in \mathcal{U}$ and denoted $f_{V,\mathcal{H}}$ is such that $f_{V,\mathcal{H}} \circ \Phi_{\mathcal{H}}$ is the characteristic function of V. In this correspondence, an expression of V as a union of pseudo-polyhedra based on \mathcal{H} coincides with a disjunctive normal form (DNF) of $f_{V,\mathcal{H}}$. For two arrangements $\mathcal{H} \subseteq \mathcal{G}$ and a function $f:D_{\mathcal{H}} \to \{0,1\}$, there is a unique function $g:D_{\mathcal{G}} \to \{0,1\}$ satisfying $f \circ \Phi_{\mathcal{H}} = g \circ \Phi_{\mathcal{G}}$. Function g is called the expansion of f from \mathcal{H} to \mathcal{G} .

Recall that a partial Boolean function $f: X \subset \{0,1\}^h \to \{0,1\}$ is threshold if there exist $\mathbf{t} \in \mathbb{R}^h$ and $t_0 \in \mathbb{R}$ such that $f(\mathbf{b}) = \operatorname{sgn}(t_0 + \mathbf{b}^{\mathsf{T}}\mathbf{t}) \ \forall \mathbf{b} \in X$. Existence of MLPs is reduced to thresholdness of functions of the type $f_{V,\mathcal{H}}$. Exploiting this transformation of the issues from geometrical terms into a combinatorial language, the development of threshold function theory will bring some light on the complexity issues addressed in the present work.

Our recognition problems can be reformulated as follows:

 $\overline{\mathcal{LP}}_2$ -Recognition: Given an arrangement \mathcal{H} and a DNF for a function $f:D_{\mathcal{H}} \to \{0,1\}$, is f threshold?

 \mathcal{LP}_2 -RECOGNITION: Given an arrangement \mathcal{H} and a DNF for a function $f: D_{\mathcal{H}} \to \{0, 1\}$, is there an arrangement $\mathcal{G} \supseteq \mathcal{H}$ such that the expansion of f from \mathcal{H} to \mathcal{G} is threshold?

 $^{^1}H \in \mathcal{H}_V$ only if $B(H) \cap B(V)$ contains an open ball of dimension n-1 in the subspace B(H), where B(X) denotes the border of X according to the usual topology on \mathbb{R}^n .

Several results established in various works on the characterization of \mathcal{LP}_2 [GC90, ZAW92, Sho93] have a simple explanation in threshold function theory. For example, Lemma 1 is a direct consequence of two facts: (i) the negation of a threshold function is threshold; (ii) the conjunction between a single variable and a threshold function is threshold.

As second example is related to the class $\mathcal{M} \supset \mathcal{LP}_2$ defined by :

$$\mathcal{M}=\{V\in\mathcal{U}\mid \text{ there is no half-space }H, \text{ and balls }B_1,B_2 \text{ in }\mathbb{R}^n, \text{ s.t.}$$
 $(H\cap B_1)\cup (H^c\cap B_2)\subset V$, and $(H\cap B_2)\cup (H^c\cap B_1)\subset V^c \}^2$

and introduced in Proposition 8 in [GC90] (or Theorem 1 in [ZAW92]) to express a necessary condition for V to be in \mathcal{LP}_2 . The condition expressed in the definition of \mathcal{M} is translated into monotonicity of Boolean function [Mur71], which is the simplest necessary condition for a function to be threshold.

In the next section, we will take advantage on results in Boolean function theory to discuss the complexity of each class of regions introduced above :

$$\mathcal{D} \subsetneq \tilde{\mathcal{D}} \subsetneq \mathcal{L}\mathcal{P}_2 \subsetneq \mathcal{M}, \qquad \tilde{\mathcal{D}} \cap \mathcal{F}\mathcal{D} \subsetneq \overline{\mathcal{L}\mathcal{P}_2} \subsetneq \mathcal{L}\mathcal{P}_2 \cap \mathcal{F}\mathcal{D}.$$

4 Intractability results

If a lot is known in threshold function theory about completely defined Boolean function, several questions remain open when partial functions are considered. For example, thresholdness of a complete function given by a monotonic³ DNF can be tested in polynomial time [PS85], but nothing is known for partial function. On the other hand, even if $V \subset \mathbb{R}^n$ based on \mathcal{H} is given by a partial function defined on the very structured domain $D_{\mathcal{H}}$, it still generalizes the case of complete Boolean function. Indeed, for any arbitrary complete function $f:\{0,1\}^h \to \{0,1\}$, there is a region $V \subset \mathbb{R}^h$ of basis $\mathcal{H} = \{H_1, \ldots, H_h\}$, where $H_i = \{x \mid x_i \geq 0\}$, such that $f = f_{V,\mathcal{H}}$. Therefore, all the questions we might ask in our setting will be at least as hard as their equivalent formulation for complete Boolean functions. In particular, since it is co-NP-Complete to decide whether an arbitrary DNF represents a threshold function, $\overline{\mathcal{LP}}_2$ -RECOGNITION is NP-Hard. The proof that checking thresholdness is co-NP can be generalized to prove the following lemma:

Lemma 2 $\overline{\mathcal{LP}}_2$ -RECOGNITION is co-NP.

For a region $V \in \tilde{\mathcal{D}}$ given by a basis \mathcal{H} and a DNF expression for $f_{V,\mathcal{H}}$, it can be shown that (i) there is a specification of V as an iterated difference of k pseudo-polyhedra with k bounded by $|\mathcal{H}|$, (ii) the equality between V and the iterated difference can be checked in polynomial time. Thus, we have:

Lemma 3 $\tilde{\mathcal{D}}$ -Recognition is NP.

Using the fact that within \mathcal{FD} , $\mathcal{D} \subset \tilde{\mathcal{D}} \subset \overline{\mathcal{LP}}_2 \subset \mathcal{LP}_2$, a single construction will be used to prove the intractability of each of these classes:

Theorem 4 \mathcal{LP}_2 -Recognition and \mathcal{D} -Recognition are NP -Hard, $\overline{\mathcal{LP}_2}$ -Recognition is co- NP -Complete, and $\tilde{\mathcal{D}}$ -Recognition is NP -Complete.

Proof: By Lemmas 2 and 3, it remains to show that the four recognition problems are *NP*-Hard. This will be done by reduction to SAT.

An instance of SAT is given as a DNF of a Boolean function $a:\{0,1\}^n \to \{0,1\}$ with the question "is a a tautology?". Consider the Euclidian space \mathbb{R}^n and the arrangement \mathcal{H} containing the following 2n closed half-spaces:

$$H_i = \{ x \mid x_i \ge 0 \}, H_{n+i} = \{ x \mid x_i \le -1 \} \forall i = 1, \dots, n.$$

 $^{^2 \}text{For } X \subset I\!\!R^n, \, X^c \text{ denotes the complement } I\!\!R^n \backslash X.$

³ A DNF is monotonic if it does not contain positive and negative occurrences of the same variable.

Let b_i denotes the Boolean variable which is the characteristic function of H_i for all i = 1, ..., 2n. Consider the region $V \subset \mathbb{R}^n$ based on \mathcal{H} defined by

$$f_{V,\mathcal{H}}(b_1,\ldots,b_{2n}) = \begin{cases} a(b_1,\ldots,b_n) & \text{if } (b_{n+1},\ldots,b_{2n}) = (0,\ldots,0), \\ 0 & \text{if } (b_{n+1},\ldots,b_{2n}) = (1,\ldots,1), \\ 1 & \text{otherwise.} \end{cases}$$
(4)

The 4 complexity results are proved if the two following statements hold:

- (i) $V \in \mathcal{D}$ if a is a tautology;
- (ii) $V \notin \mathcal{LP}_2$ if $a^{-1}(0) \neq \emptyset$. a(b) = 0.

The proof of (i) is easy, since if a is a tautology, $\mathbf{b} = (0^n, 1^n)$ is the only point for which $f_{V,\mathcal{H}}$ is 0. Thus, $V = \mathbb{R}^n \setminus \bigcap_{i=1}^n H_{n+i}$, which is clearly in \mathcal{D} .

The proof of (ii) goes as follows. Assume that a is not a tautology and let distinguish two cases. If $\exists \mathbf{b} \in a^{-1}(0)$, with $b_j = 0$ for at least one j, we can show that $V \notin \mathcal{M}$. On the other hand, if $a^{-1}(0) = \{(1, \ldots, 1)\}$, we have to show that the region $\mathbb{R}^n \setminus (\bigcap_{i=1}^n H_i \cup \bigcap_{i=1}^n H_{n+i})$ is not in \mathcal{LP}_2 . This situation is a generalization to n dimensions of the example pictured in Figure 5 in [Gib96] and the proof is based on another property of threshold function known as assumability [Mur71].

Deciding whether a Boolean function given by an arbitrary DNF is monotonic or not is NP-Complete. From this we can establish the following result:

Proposition 5 Given a region V specified by a basis \mathcal{H} and its expression as a union of intersections of elements of \mathcal{H} , deciding whether $V \in \mathcal{M}$ is NP-Complete.

5 Tractable cases

The complexity results of Theorem 4 rely essentially on the fact that it is hard to find a DNF expression characterizing the complement of a region defined by a DNF, since it is already hard to decide whether this complement is empty or not (SAT). The first tractable cases listed in this section consist of situations where both V and V^c are available as input.

Proposition 6 If V is given by a basis \mathcal{H} and two DNFs, one for $f_{V,\mathcal{H}}$ and the other for $f_{V^c,\mathcal{H}}$, then it can be decided in polynomial time whether V is in \mathcal{D} (similarly in $\tilde{\mathcal{D}}$).

The similar question about $\overline{\mathcal{LP}}_2$ was solved only with an additional assumption :

Proposition 7 If V is given by a basis \mathcal{H} and two DNFs D_1 and D_2 for $f_{V,\mathcal{H}}$ and $f_{V^c,\mathcal{H}}$ such that $D_1(\mathbf{b}) = D_2(\mathbf{b}) = 0 \ \forall \mathbf{b} \notin D_{\mathcal{H}}$, then it can be decided in polynomial time whether V is in $\overline{\mathcal{LP}}_2$.

In many practical situations, the size n of the Euclidian space is small, while the number of half-spaces can be very large. Therefore, it is worth highlighting that some of the recognition problems become easy when n is a constant.

Proposition 8 $\overline{\mathcal{LP}}_2$ -Recognition, \mathcal{D} -Recognition and \mathcal{D} -Recognition are polynomial in the size of the basis and in the size of the DNF given as input.

6 Open questions

It can be decided in polynomial time whether a Boolean function given by a monotonic DNF is threshold. Thus, the intractability of $\overline{\mathcal{LP}}_2$ -Recognition might be connected to the intractability of class \mathcal{M} (Proposition 5). We addressed the same question as in $\overline{\mathcal{LP}}_2$ -Recognition with the additional

assumption that the inputs are given in a monotonic form, *i.e.* the basis does not contain two half-spaces complement of each other and the DNF is monotonic. The complexity of this question is still open.

As mentioned in section 2, there exist some regions in \mathcal{LP}_2 for which all the known expressions of their characteristic functions in the form of equations (1-3) require m exponential in the size of the input. An important and challenging problem would be to demonstrate some lower bounds on m which could be exponential or even super-polynomial in the size of the input.

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