

Reliable Learning Using Post Classes

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ABSTRACT

The complexity of the consistency problem for several important classes of Boolean functions is analyzed. The classes of functions under investigation are those which are closed under function composition or superposition. Several of these so called Post classes are considered within the context of machine learning with an application to breast cancer diagnosis. The considered Post classes furnish a user-selectable measure of reliability. It is shown that for realistic situations which may arise in practice, the consistency problem for these classes of functions is polynomial-time solvable.

1 Introduction

The consistency problem is an important problem in computational learning theory [2], [13] and can be thought of as a search of a rule from examples. That is, given some sets T and F of “true” and “false” vectors, respectively, we aim to discover a Boolean function f which takes on the value 1 for all vectors in T and the value 0 for all vectors in F . We may also assume that the target function f is chosen from some class of possible target functions. One important reason for studying the complexity of the consistency problem is its relation to the PAC approximate learning model of Valiant [13]. If the consistency problem for a given class is NP-hard, then this class is not PAC-learnable. Moreover, this would also imply that this class cannot be learned with equivalence queries [1].

In practice, it is often reasonable to require f to possess certain properties, known a priori, or equivalently, to belong to some specific class of functions. For example, the class of monotone Boolean functions is often representative of real-life phenomena and has been studied in the context of learning in such fields as medical diagnosis [6], manufacturing and reliability analysis [7], as well as signal processing [15]. In [3], the consistency problem was considered for various classes of Boolean functions, such as positive (monotone), k -DNF, h -term-DNF, self-dual, and many others. Specifically, the question of the existence of polynomial-time algorithms for various classes of functions was investigated. In this

paper, we consider the consistency problem for several important classes of Boolean functions, known as Post classes.

In 1921, the American mathematician Emil Post described the structure of all classes of Boolean functions which are closed under superposition [11]. The results of Post have been applied outside the field of algebra of logic, for example, in the synthesis [10] and reliability [9] of control systems. For example, the class M of monotone Boolean functions is a Post class, since a superposition of monotone Boolean functions results in a monotone Boolean function. The consistency problem for M was shown to be polynomial-time solvable in [3]. In fact, this was shown to be the case for all transitive classes. In this paper, the classes under consideration are F_8^μ , F_4^μ , F_7^μ , F_3^μ and F_8^∞ , F_4^∞ , F_7^∞ , F_3^∞ , using the notation of Yablonsky [14]. This and other notation is established in Section 2. The above classes have practical relevance in the context of learning. This is discussed in Section 3 where an example of breast cancer diagnosis is considered. Finally, Section 4 contains the main results concerning the complexity of the consistency problem for these classes of Boolean functions.

2 Definitions and notation

Let E^n denote the n -cube $\{0, 1\}^n$. A Boolean function is a mapping $f : E^n \rightarrow E^1$. The elements $v \in E^n$ will be referred to as vectors. The relation $\alpha \preceq \beta$ holds for two vectors $\alpha = (\alpha_1, \dots, \alpha_n)$ and $\beta = (\beta_1, \dots, \beta_n)$ if $\alpha_1 \leq \beta_1, \dots, \alpha_n \leq \beta_n$. A Boolean function f is monotone (also called positive) if for any two vectors α and β such that $\alpha \preceq \beta$, the inequality $f(\alpha) \leq f(\beta)$ holds. The notation for various Post classes chosen here coincides with that used by Post as well as that in [14].

We now proceed to define several important Post classes.

Definition 1 *A function f satisfies condition $\langle A^\mu \rangle$ (resp. $\langle a^\mu \rangle$), $\mu \geq 2$, if for any collection of μ vectors $v^{(1)}, v^{(2)}, \dots, v^{(\mu)} \in E^n$ for which $f(v^{(1)}) = \dots = f(v^{(\mu)}) = 1$ (resp. $f(v^{(1)}) = \dots = f(v^{(\mu)}) = 0$), there exists an $i \in \{1, \dots, n\}$ such that $v_i^{(1)} = v_i^{(2)} = \dots =$*

$$v_i^{(\mu)} = 1 \text{ (resp. } v_i^{(1)} = v_i^{(2)} = \dots = v_i^{(\mu)} = 0).$$

Let $T(f) = \{v \in E^n : f(v) = 1\}$ be called the on-set of function f . Similarly, let $F(f) = \{v \in E^n : f(v) = 0\}$ be the off-set of f .

Definition 2 A function f satisfies condition $\langle A^\infty \rangle$ (resp. $\langle a^\infty \rangle$) if there exists an $i \in \{1, \dots, n\}$ such that $v_i = 1$ (resp. $v_i = 0$) for all vectors $v \in T(f)$, (resp. $v \in F(f)$).

It is clear that if a function satisfies $\langle A^{\mu_1} \rangle$, then it also satisfies $\langle A^{\mu_2} \rangle$ for $2 \leq \mu_2 \leq \mu_1$. Also, the condition $\langle A^\infty \rangle$ implies $\langle A^\mu \rangle$ for any $\mu \geq 2$. Similar statements can be made for $\langle a^\mu \rangle$ and $\langle a^\infty \rangle$. Let F_8^μ and F_4^μ be the sets of all functions satisfying conditions $\langle A^\mu \rangle$ and $\langle a^\mu \rangle$, respectively. Similarly, let F_8^∞ and F_4^∞ be the set of functions satisfying $\langle A^\infty \rangle$ and $\langle a^\infty \rangle$. Finally, let F_7^μ (resp. F_3^μ) and F_7^∞ (resp. F_3^∞) be the classes of monotone functions satisfying conditions $\langle A^\mu \rangle$ (resp. $\langle a^\mu \rangle$) and $\langle A^\infty \rangle$ (resp. $\langle a^\infty \rangle$).

Given sets $T \subseteq E^n$ and $F \subseteq E^n$, such that $T \cap F = \emptyset$, the partially defined function $g_{T,F}$ is defined as

$$g_{T,F}(v) = \begin{cases} 1, & v \in T \\ 0, & v \in F \\ *, & \text{otherwise} \end{cases}$$

A function f is called an *extension* of $g_{T,F}$ if $T \subseteq T(f)$ and $F \subseteq F(f)$ [3]. The consistency problem (also called the extension problem) can be posed as: given some class C of functions and two sets T and F , is there an extension $f \in C$ of $g_{T,F}$? The problem for class C will be denoted as $\text{EXTENSION}(C)$ [3] and the question of its polynomial-time solvability is central in the PAC theory of learning.

3 Post classes and breast cancer diagnosis

In [8], a breast cancer diagnosis problem is considered within a learning framework. The goal is to find a set of rules, using some specific set of features, to determine whether or not an individual case is highly suspicious for malignancy. Another goal may be to determine whether or not a biopsy or short term follow-up is necessary. A radiologist studying a breast tumor may use a number of different features or attributes to make the decision regarding malignancy or biopsy. Some such attributes might be: number of calcifications per cm^2 , total number of calcifications, variation in the shape of the calcifications, approximate volume of the lesion, and many others [16].

Each such attribute can be encoded by an appropriate number of binary valued variables. For example, suppose the total number of calcifications can be A: < 10 , B: 10 to 30, or C: > 30 . Thus, two binary variables would be needed to represent each case, e.g. (0,0) for A, (0,1) for B, and (1,0) for C. Accordingly, a binary

vector $x = (x_1, \dots, x_n)$ of the necessary length to encode all attributes represents any given breast tumor case. The goal then becomes to find a Boolean function f that would classify each example vector x according to the rules that are used by the expert, or radiologist. For example, $f(x) = 1$, if the tumor encoded by x is believed to be malignant, and $f(x) = 0$, otherwise. This can be considered to be a form of knowledge discovery from examples. In [8], it was assumed that the functions underlying the patterns are monotone, or in other words, $f \in M$. One reason for this seemingly restrictive assumption is rooted in the fact that any Boolean function can be expressed in terms of several increasing and decreasing monotone Boolean functions [12]. Another reason is that the direction of the variables that cause the phenomenon to occur or indicate its presence is known (i.e. $x_i = 1$ "causes" the phenomenon and $x_i = 0$ doesn't). In that case, the polarities of variables can be changed appropriately so as to make the function monotone without loss of any generality.

With the intention of adding a certain degree of reliability, it is natural to expect, or rather, demand, that a given number of positive examples, that is, those which are classified as being malignant, should exhibit some agreement or concord among their attributes. We note here that what is meant by the term *reliability* is not the traditional *confidence* in the PAC learning sense. Accordingly, reliability does not refer to the process of learning itself, but rather to the constraints imposed on the hypothesis space.

The weakest such requirement of reliability is that any two malignant cases must necessarily agree on at least one attribute, be it the total number of calcifications, approximate volume of the lesion, or any other. Of course, this attribute can be different depending on the two examples selected. For instance, malignant cases M1 and M2 might agree on attribute A, whereas malignant cases M3 and M4 might agree on attribute B, and so on. One way to strengthen this requirement is to increase the number μ of positive examples all of which must possess one or more common attributes. That is, any μ malignant cases must share at least one common attribute. Such a condition is equivalent to restricting the function f to be in the class F_8^μ or F_7^μ , if f is additionally assumed to be monotone. The strongest, and perhaps unreasonably restrictive, versions of this notion of reliability are captured by the classes F_8^∞ and F_7^∞ , which call for *all* malignant cases to agree on at least one attribute.

The restriction to such classes, however, necessitates the use of more bits for the encoding of the cases. That is, one bit is required for every value of an attribute. So, for example, an attribute that can take three values A: < 10 , B: 10 to 30, or C: > 30 should be coded as (1,0,0), (0,1,0), and (0,0,1) respectively. Such an encoding scheme would then allow one to check whether or not two or more malignant cases agree on the value

of any given attribute. Having discussed the possible practical aspects of the above mentioned Post classes in the context of learning theory, we now move on to the main results in which we consider the complexity of the consistency problem for these classes.

4 Extensions in F_8 , F_7 , F_4 , and F_3

We first consider the problems $\text{EXTENSION}(F_8^\mu)$ and $\text{EXTENSION}(F_7^\mu)$ for a fixed constant μ . The results for F_4^μ and F_3^μ are identical and can be obtained using the duality principle. The question of polynomial-time solvability is posed in terms of n , $|T|$, and $|F|$.

Theorem 3 *EXTENSION(F_8^μ) and EXTENSION(F_7^μ) are polynomial-time solvable for a fixed constant μ .*

Proof. For the case of F_8^μ , we first must show that checking whether there exists an $i \in \{1, \dots, n\}$ such that $v_i^{(1)} = v_i^{(2)} = \dots = v_i^{(\mu)} = 1$ for a collection of μ vectors, takes time $O(\mu \cdot n)$. This is because we require n passes through all μ vectors, once for each coordinate, keeping a flag set if the coordinate is equal to 1. This condition is equivalent to saying that the conjunction of all μ vectors has non-zero Hamming weight.

For a fixed constant μ , checking the above condition for all possible μ -collections takes $O(\mu \cdot n \cdot |T|^\mu)$ time, which is polynomial in $|T|$ and n . If this condition is satisfied for all μ -collections of vectors from T , then as an extension in F_8^μ , we provide

$$f(v) = \begin{cases} 1, & v \in T \\ 0, & \text{otherwise} \end{cases} \quad (1)$$

For the case of F_7^μ , the above steps must be preceded by a check for monotonicity. This entails checking whether there is no pair of vectors $v \in T$ and $u \in F$ such that $v \preceq u$, which takes $O(n \cdot |T| \cdot |F|)$ time. Only if this condition is satisfied can an extension in F_7^μ possibly exist. Thus, $\text{EXTENSION}(F_8^\mu)$ and $\text{EXTENSION}(F_7^\mu)$ are both polynomial-time solvable in n , $|T|$, and $|F|$. ■

We now examine the case in general, when $\mu \leq |T|$ is part of the input to the problems $\text{EXTENSION}(F_8^\mu)$ and $\text{EXTENSION}(F_7^\mu)$.

Theorem 4 *EXTENSION(F_8^μ) and EXTENSION(F_7^μ) are co-NP-complete.*

Proof. The problem $\text{EXTENSION}(F_8^\mu)$ has a negative answer if there exists an η -collection of vectors $V \subseteq T$, ($|V| = \eta$ and $\eta \leq \mu$), such that for all coordinates $i \in \{1, \dots, n\}$, there is at least one vector $v \in V$ for which $v_i = 0$. Thus, it is easy to see that $\text{EXTENSION}(F_8^\mu) \in \text{co-NP}$, since a nondeterministic algorithm need only guess the correct collection of η vectors $v^{(1)}, \dots, v^{(\eta)} \in T$ that satisfy the above property, and check this fact in polynomial time as in Theorem 3.

We show that $\text{EXTENSION}(F_8^\mu)$ is co-NP-hard by a reduction from the well-known MINIMUM-COVER

problem [4], which is known to be NP-complete. The instance of the MINIMUM-COVER problem is a collection C of subsets of a finite set S and a positive integer $\mu \leq |C|$. The question is: does C contain a cover for S of size μ or less, that is, a subset $C' \subseteq C$ with $|C'| \leq \mu$ such that every element of S belongs to at least one member of C' ?

Consider a finite set $S = \{x_1, \dots, x_n\}$ and an arbitrary collection of subsets $C \subseteq 2^S$. Let us associate a set T to C by defining

$$T = \left\{ \overline{(v^c)} : c \in C \right\}$$

where \overline{v} denotes the bit-wise complement of vector v and v^c is the characteristic vector of set $c \in C$, that is, $x_i \in c \iff v_i^c = 1$. This set can be constructed from C in polynomial time in n and $|C|$. It is easy to see that C contains a cover for S if and only if $\text{EXTENSION}(F_8^\mu)$ has a negative answer.

Turning to F_7^μ , we note that the added constraint of monotonicity does not make the problem any easier. We have seen that the ‘‘hardness’’ of the problem lies entirely within the set T of true vectors, whereas monotonicity is concerned with the relationship between the sets T and F . So, even if we were furnished with a guarantee of monotonicity a priori, we would still have to solve the same exact problem as above, namely, to check whether condition $\langle A^\mu \rangle$ is satisfied within set T . Thus, we can conclude that $\text{EXTENSION}(F_7^\mu)$ is also co-NP-complete. ■

Now, let us turn to the problem of finding an extension in F_8^∞ and F_7^∞ , with F_4^∞ and F_3^∞ again being identical by duality. It is easy to see that checking whether all vectors $v \in T$ have at least one common unit component takes $O(n \cdot |T|)$ time. The question of monotonicity can be answered beforehand in $O(n \cdot |T| \cdot |F|)$ time, as in Theorem 3. In that case, an extension in F_8^∞ or F_7^∞ can be constructed as in (1), in polynomial time, giving us the following result.

Theorem 5 *EXTENSION(F_8^∞) and EXTENSION(F_7^∞) are polynomial-time solvable.*

Having considered the complexity of the consistency problem for the above Post classes, it is worthwhile to ask just how restrictive, in the cardinality sense, are the hypotheses spaces (F_8^μ, F_8^∞) and (F_7^μ, F_7^∞) , relative to the classes of all Boolean functions and monotone functions, respectively. After all, if the imposed requirement of reliability, the degree of which is controlled by μ , would significantly restrict the hypothesis space, then their use in a practical setting would be questionable. Partial answers to these questions can be found in the work of Korshunov [5], who considered asymptotic formulae for the cardinality of the above classes of functions. Due to space limitations, we will not attempt to reproduce all the asymptotic results,

many of which are different for even and odd n , and refer the interested reader to the original paper. However, it is worth mentioning that the cardinalities of the classes of functions considered here grow very quickly as $n \rightarrow \infty$. For example, for any fixed $\mu \geq 3$ and $n \rightarrow \infty$, $|F_8^\mu| \sim |F_8^\infty| \sim n2^{2^{n-1}}$, while for $\mu = 2$,

$$\begin{aligned} |F_8^2| &> 2^{2^{n-1}} (3/2)^{\binom{n-1}{2}}, & \text{for odd } n \geq 3 \\ |F_8^2| &> 2^{2^{n-1}} (3/2)^{\frac{1}{2}\binom{n}{2}}, & \text{for even } n > 3. \end{aligned}$$

Similar asymptotic relationships exist between (F_7^μ, F_7^∞) and the cardinality of the class of monotone functions. These facts are somewhat reassuring in that the imposed reliability constraints do not seem to lead to an overly restrictive hypothesis space.

5 Conclusion

We have analyzed the complexity of the consistency problem for several important Post classes, which have shown to be relevant in the study of reliability of control systems. Additionally, we discussed a possible application of the considered classes to diagnosis of breast cancer. The positive result obtained here is that if μ is known a priori and fixed, which would reflect the case in practice, the consistency problem in classes $F_8^\mu, F_4^\mu, F_7^\mu$ and F_3^μ is solvable in polynomial time. In the context of breast cancer diagnosis, the parameter μ would reflect the user-settable degree of required reliability for the rules being inferred. The higher the value of μ , the more strict this requirement becomes.

Additionally, the consistency problem for classes $F_8^\infty, F_4^\infty, F_7^\infty$ and F_3^∞ is also polynomial time solvable. As part of future work, it would be worthwhile to consider best-fit extensions in the above classes of Boolean functions, especially for high values of μ , when a fewer number of possible rules is expected.

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