

# LEARNING PARTIAL BOOLEAN FUNCTIONS BY LOCAL SEARCH

BJOERN ANDRES

ABSTRACT. We seek to separate disjoint sets  $A \cup B = S \subseteq \{0, 1\}^J$  by Boolean functions  $f, g: \{0, 1\}^J \rightarrow \{0, 1\}$  such that  $f(A) = 1$  and  $g(B) = 1$  and  $f \cdot g = 0$ . Specifically, we consider  $f$  and  $g$  each defined by a disjunctive normal form, and we seek to minimize the sum of the lengths of these forms. This problem is known to be inapproximable to within  $(1 - \epsilon) \log |S|$  for any  $\epsilon > 0$  unless  $P=NP$ . The task of this project is to design, implement and compare local search algorithms for the problem.

## 1. PRELIMINARIES

**Definition 1.** A tuple  $(J, S, X, Y, x, y)$  is called *Boolean labeled data* iff  $J$  is finite and non-empty, called the set of *features*,  $X = \{0, 1\}^J$ , called the *feature space*,  $S$  is finite and non-empty, called the set of *samples*,  $Y = \{0, 1\}$ , called the *label space*,  $x: S \rightarrow X$  and  $y: S \rightarrow Y$ . Boolean labeled data is called *ambiguous* iff there exist  $s, s' \in S$  such that  $x_s = x_{s'}$  and  $y_s \neq y_{s'}$ .

**Definition 2.** For any finite  $X = \{0, 1\}^J$ ,  $\Gamma = \{(V, \bar{V}) \in 2^J \times 2^J \mid V \cap \bar{V} = \emptyset\}$  and  $\Theta = 2^\Gamma$ , the family  $f: \Theta \rightarrow \{0, 1\}^X$  such that, for any  $\theta \in \Theta$  and any  $x \in X$ ,

$$f_\theta(x) = \sum_{(J_0, J_1) \in \theta} \prod_{j \in J_0} x(j) \prod_{j \in J_1} (1 - x(j)) \quad (1)$$

is called the family of  $J$ -variate *disjunctive normal forms (DNFs)*.

For  $R_l, R_d: \Theta \rightarrow \mathbb{N}_0$  such that for all  $\theta \in \Theta$ ,

$$R_l(\theta) = \sum_{(J_0, J_1) \in \theta} (|J_0| + |J_1|) \quad , \quad (2)$$

$$R_d(\theta) = \max_{(J_0, J_1) \in \theta} (|J_0| + |J_1|) \quad , \quad (3)$$

$R_l(\theta)$  and  $R_d(\theta)$  are called the *length* and *depth*, resp., of the DNF defined by  $\theta$ .

**Definition 3.** For any non-ambiguous Boolean labeled data  $D = (J, S, X, Y, x, y)$ , for any non-empty set  $\Theta$  and family  $f: \Theta \rightarrow \{0, 1\}^X$  of Boolean functions, and

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TU DRESDEN, UNIVERSITY OF TÜBINGEN  
E-mail address: bjoern.andres@tu-dresden.de.

for any function  $R: \Theta \rightarrow \mathbb{N}_0$ , called a *regularizer*, the instance of the OPTIMAL-SEPARATION problem w.r.t.  $D, \Theta, f$  and  $R$  has the form

$$\min_{(\theta, \theta') \in \Theta^2} R(\theta) + R(\theta') \quad (4)$$

$$\text{subject to } \forall s \in y^{-1}(1): f_\theta(x_s) = 1 \quad (\text{exactness}) \quad (5)$$

$$\forall s \in y^{-1}(0): f_{\theta'}(x_s) = 1 \quad (\text{exactness}) \quad (6)$$

$$f_\theta \cdot f_{\theta'} = 0 \quad (\text{non-contradictoriness}) \quad (7)$$

For any  $m \in \mathbb{N}_0$ , the instance of the SEPARABILITY problem w.r.t.  $D, \Theta, f, R$  and  $m$  is to decide whether there exist  $\theta, \theta' \in \Theta$  such that

$$\forall s \in y^{-1}(1): f_\theta(x_s) = 1 \quad (\text{exactness}) \quad (8)$$

$$\forall s \in y^{-1}(0): f_{\theta'}(x_s) = 1 \quad (\text{exactness}) \quad (9)$$

$$f_\theta \cdot f_{\theta'} = 0 \quad (\text{non-contradictoriness}) \quad (10)$$

$$R(\theta) + R(\theta') \leq m \quad (\text{boundedness}) \quad (11)$$

Two remarks are in order. Firstly, any feasible solution  $(\theta, \theta') \in \Theta^2$  to either of these problems defines a partial Boolean function  $h: X' \rightarrow \{0, 1\}$  with  $X' = \{x' \in \{0, 1\}^J \mid f_\theta(x') = 1 \vee f_{\theta'}(x') = 1\}$  and such that for all  $x' \in X'$ , we have  $h(x') = 1$  iff  $f_\theta(x') = 1$  (and equivalently,  $h(x') = 0$  iff  $f_{\theta'}(x') = 1$ ). Secondly, exactness means the labeled data is classified totally and without errors, i.e., for all  $s \in S$ , we have  $x_s \in X'$  and  $h(x_s) = y_s$ .

## 2. MOTIVATION

SEPARABILITY is known to be NP-hard, and OPTIMAL-SEPARABILITY is known to be inapproximable to within  $(1 - \epsilon) \log |S|$  for any  $\epsilon > 0$  unless  $P=NP$ . It is unknown how close one can get to a  $\log |S|$ -approximation in polynomial time.

## 3. RELATED WORK

The standard textbook on Boolean functions is [2], and its two authors are leading researchers in this area. References to their publications are excellent pointers to specific recent work.

## 4. TASK

**The task of this project is to design, implement and empirically compare local search algorithms for the problem:**

- (1) A feasible solution to begin (initialize) with is given by

$$f_\theta(x) = \sum_{s \in y^{-1}(1)} \prod_{\{j \in J \mid x_s(j)=1\}} x(j) \prod_{\{j \in J \mid x_s(j)=0\}} (1 - x(j)) \quad (12)$$

$$f_{\theta'}(x) = \sum_{s \in y^{-1}(0)} \prod_{\{j \in J \mid x_s(j)=1\}} x(j) \prod_{\{j \in J \mid x_s(j)=0\}} (1 - x(j)) \quad (13)$$

Others can be proposed as part of this project.

- (2) Two transformations that shall be searched over in order to reduce the objective function are described below. Others are to be proposed as part of this project.
  - (a) removing a literal from a term

- (b) adding a literal to one term and simultaneously removing literals of the same variable from other terms.
- (3) Algorithms are to be compared in terms of the objective value with respect to the running time, for any binary classification problem, e.g., distinguishing the digits 1 and 7 in the MNIST data set of hand-written digits [1].

## REFERENCES

- [1] <http://yann.lecun.com/exdb/mnist/>.
- [2] Yves Crama and Peter L. Hammer. *Boolean Functions: Theory, Algorithms, and Applications*. Cambridge University Press, 2011.