

# THE REVERSE ISING PROBLEM

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## 1. INTRODUCTION AND TERMINOLOGY

**Definition 1.1.** Let  $\Sigma = \{-1, +1\}$ . An **Ising circuit** is a function  $f : \Sigma^N \rightarrow \Sigma^M$  where  $N$  and  $M$  are finite subsets of  $\mathbb{N}$ . For convenience we always assume  $N = \{1, \dots, n\}$  and  $M = \{n+1, \dots, n+m\}$ .

## 2. HIERARCHICAL CLUSTERING

A choice of “auxiliary array” can be thought of as a function  $a : \Sigma^N \rightarrow \Sigma^A$  which assigns an auxiliary state to each input state. The collection of germs of this function partitions  $\Sigma^N$  into subsets of inputs which share the same auxiliary state. To be clear, we’re talking about the partition

$$\{a^{-1}(\alpha)\}_{\alpha \in \Sigma^A}$$

of input spin space. If the choice of auxiliary array  $a$  makes the Ising circuit feasible, then we say  $a$  solves the Ising circuit.

This simple observation motivates a simple question: *can a partition of  $\Sigma^N$  be found such that it matches the partition produced by some feasible auxiliary array using only the logic of the Ising circuit?* Producing such a partition is a clustering problem on input spinspace.

## 3. PSEUDO-BOOLEAN OPTIMIZATION AND POLYNOMIAL FITTING

A pseudo boolean function (PBF) is any function  $f : \{0, 1\} \rightarrow \mathbb{R}$ . It is a well known fact that any such PBF can be uniquely represented by a multilinear polynomial in  $n$  variables [pseudo-boolean optimization Boros, Hammer]; that is, a polynomial

$$g(x_1, \dots, x_n) = \sum_{S \subseteq [n]} a_S \prod_{j \in S} x_j$$

with  $a_S \in \mathbb{R}$  which equals  $f$  pointwise on  $\{0, 1\}^n$ . To be clear, here  $S$  iterates over all subsets of  $[n] = \{1, \dots, n\}$ .

It is another well-known fact that the optimization of any pseudo-boolean function can be reduced in polynomial time to an optimization problem on a quadratic polynomial. The original method for accomplishing this was first written by Rosenberg, and since then a reputable zoo of alternative algorithms have been introduced. Most methods share the same basic idea: reduce degree  $\geq 3$  monomial terms appearing in the polynomial  $g$  by introducing auxiliary variables subject to constraints.

<copy Rosenberg algorithm from Boros, Hammer pg 168>

**Theorem 3.1.** *Let  $f$  be a multilinear polynomial in  $n$  variables. There exists an algorithm REDUCE which produces a multilinear polynomial  $g$  in  $n + a$  variables such that*

$$\min_{(\mathbf{x}, \mathbf{a}) \in \mathbb{B}^n \times \mathbb{B}^a} g(\mathbf{x}, \mathbf{a}) = \min_{\mathbf{x} \in \mathbb{B}^n} f(\mathbf{x})$$

and if  $(\mathbf{x}, \mathbf{a}) = \arg \min_{(\mathbf{x}, \mathbf{a}) \in \mathbb{B}^n \times \mathbb{B}^a} g(\mathbf{x}, \mathbf{a})$  then  $\mathbf{x} = \arg \min_{\mathbf{x} \in \mathbb{B}^n} f(\mathbf{x})$ .

**Boros Hammer Pseudo Boolean Optimization 2002.**

□

We need a slightly stronger statement however.

**Theorem 3.2.** *Let  $f : \Sigma^N \rightarrow \Sigma^M$  be a circuit. Then there exists an Ising circuit with auxiliary spins given by Hamiltonian  $H$  which solves  $f$ .*

**Proof.** Fix  $G = N \cup M$  and consider the hamming objective function  $\text{ham} : \Sigma^N \times \Sigma^M \rightarrow \mathbb{R}$  defined

$$\text{ham}(s, t) = d(t, f(s))$$

where  $d(t, f(s))$  is the Hamming distance between  $t$  and the correct output  $f(s)$ . Then there exists some multilinear polynomial  $g$  in  $|G|$  variables which recovers  $\text{ham}$  pointwise. We now apply Rosenberg reduction to  $g$  and set  $H$  equal to the terminal quadratic polynomial we obtain. All that remains to show is that on any input level  $s$  the output which minimizes  $H$  is  $f(s)$ .

Fix an input  $s$  and suppose that the minimizer of  $g^k(s, \cdot)$  has output coordinates  $f(s)$ . To obtain  $g^{k+1}$  we replace some pair  $x_i x_j$  by  $x_{k+1}$  and add the expression  $M(x_i x_j - 2x_i x_{k+1} - 2x_j x_{k+1} + 3x_{k+1})$ . Observe that this expression is zero if  $x_i x_j = x_{k+1}$  and is strictly positive otherwise. It follows that  $g^k(\mathbf{x}) = g^{k+1}(\mathbf{x}, x_{k+1})$  if  $x_{k+1} = x_i x_j$  and  $g^k(\mathbf{x}) < g^{k+1}(\mathbf{x}, x_{k+1})$  if  $x_{k+1} \neq x_i x_j$ . Hence the minimizer of  $g^{k+1}$  on input level  $s$  also has the correct output coordinates, and inductively, we conclude that  $H$  is an Ising Hamiltonian reproducing the circuit  $f$ .  $\square$