# Project Weak-Schur: Ising Model Proposal

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# 1 Hamiltonian Function 1

We are searching for a colouring of the first N natural numbers into n colours, such that whenever  $\exists (a,b,c) \in [N]$  and a+b=c, then a,b and c do not share the same colour. The following have been modelled after the decision variables and functions presented in [1], Section 6.

**Definition 1.** The decision variable  $x_{m,c}$  is defined as

$$x_{m,c} := \begin{cases} 1 & Col(m) = c \\ 0 & else \end{cases} \tag{1}$$

This defines the matrix of decision variables  $\mathbf{x} \in M_{N \times n}(\{0,1\})$ .

**Definition 2.** The Hamiltonian function for a Schur partition of the first N natural numbers into n colors is the following.

$$\mathcal{H}_{S}(\boldsymbol{x}) := A \sum_{i=1}^{N} \left( 1 - \sum_{c=1}^{n} x_{i,c} \right)^{2} + B \sum_{i,j=1}^{N} \sum_{c=1}^{n} x_{i,c} x_{j,c} x_{i+j,c}$$
(2)

where  $A, B \in \mathbb{R}_{>0}$  are arbitrary weights.

We now add a constraint to define the Hamiltonian for the Weak Schur case.

**Definition 3.** The Hamiltonian function for a Weak Schur partition of the first N natural numbers is the following.

$$\mathcal{H}_{WS}(\boldsymbol{x}) = A \sum_{i=1}^{N} \left( 1 - \sum_{c=1}^{n} x_{i,c} \right)^{2} + A$$

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$$\underbrace{B\sum_{i,j=1}^{N}\sum_{c=1}^{n}x_{i,c}x_{j,c}x_{i+j,c}\cdot(2N-|i-j|)}_{\mathcal{B}}$$
 (4)

where  $A, B \in \mathbb{R}_{\geq 0}$  are arbitrary weights.

#### 1.1 Remarks on Definitions

We make some remarks on the first sum A first.

1.  $\mathcal{A}$  penalises the assignment of multiple colors to a single number.

Now, we study  $\mathcal{B}$ .

- 1. Notice that for any good triplet, (i, j, k),  $l_{i,j} = x_{i,c}x_{j,c}x_{i+j,c} = 0$ , we have  $x_{i,c} = 0$ , or  $x_{i+j,c} = 0$ , without loss of generality, which immediately implies that they do not share the colour c.
- 2. Suppose that they all share the same colour, then,  $l_{i,j} = 1$ , and the energy penalty becomes positive in  $\mathcal{B}$ .
- 3. In the degenerate case when i = j, |i j| = 0, and  $\mathcal{B}$  is effectively maximised by a multiplication with 2N, even if it satisfies the sum-free constraints. This implies only weakly sum-free triplets can attain a minimal  $\mathcal{B}$ . Intuitively,  $\mathcal{B}$  is minimised whenever i and j are "far apart".

#### 1.2 Problem Visualisation

The problem is essentially an n-dimensional Ising problem with N charges.

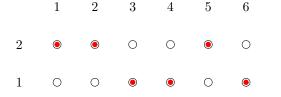


Figure 1: Ising Model Representation of the partition  $\{\{1,2,5\},\{3,4,6\}\}$  into n=2 colours. Here, Red = +1, White = 0.

### 1.3 Computational Aspects

To simplify the computation of the Hamiltonian, we can use the following looping order:

$$\mathcal{B}_{WS}(\boldsymbol{x}) = B \sum_{c=1}^{n} \sum_{i=1}^{N} x_{i,c} \sum_{j=1}^{N} x_{j,c} x_{i+j,c} (2N - |i-j|)$$

A simple Simulated Annealing algorithm can be proposed to find minimal configurations<sup>1</sup> for a given N, n, similarly to a simulated annealing approach, as follows in Alg. 1. Other approaches can be implemented from [2], [3], [4] and [5].

Algorithm 1: Simulated Annealing solution

$$\begin{array}{l} \mathbf{Data:} \ T_{max}, T_{min} \geq 0, \ N, n \in \mathbb{N} \\ \mathbf{Result:} \ x \in M_{N \times n}(\{0,1\}) \\ x \overset{\$}{\leftarrow} \ M_{N \times n}(\{0,1\}); \\ T \leftarrow T_{max}; \\ \mathbf{while} \ T > T_{min} \ \mathbf{do} \\ & \left| \begin{array}{c} r, c \overset{\$}{\leftarrow} [1:N] \times [1:n]; \\ x_{r,c}^{\mathrm{new}} \leftarrow -x_{r,c}; \\ \mathbf{if} \ \mathcal{H}_{WS}(x^{new}) \leq \mathcal{H}_{WS}(x) \ \mathbf{then} \\ & | \ x \leftarrow x^{\mathrm{new}}; \\ \mathbf{else} \\ & \left| \begin{array}{c} \mathbf{if} \ e^{-\frac{\mathcal{H}_{WS}(x^{new}) - \mathcal{H}_{WS}(x)}{T}} < \alpha \ \mathbf{then} \\ & | \ x \leftarrow x^{\mathrm{new}}; \\ & | \ \mathbf{end} \\ & | \ \mathbf{end} \\ \end{array} \right. \end{array}$$

## 1.4 Optimisation Formulation

Alternatively, we can formulate the problem as a constrained, non-linear, integer optimisation problem in the following way.

$$\min_{\boldsymbol{x} \in \{0,1\}^N} H_{WS}(\boldsymbol{x}) \tag{5}$$

Using the constraints, we get that

$$\min_{\boldsymbol{x} \in \{0,1\}^N} \mathcal{B}(\boldsymbol{x})$$
 subject to:  $\forall c, \ \langle \mathbb{1}, x_{\cdot,c} \rangle = 1$ 

or, considering  $x \in M_{N \times n}(\{0,1\}),$ 

$$\min_{\boldsymbol{x} \in \{1,-1\}^N} \mathcal{B}(\boldsymbol{x}) \tag{6}$$

subject to: 
$$\langle \mathbb{1}, \boldsymbol{x} \rangle = \mathbb{1}$$
 (7)

### 1.5 Decoding

Here, the decoding is straightforward: A valid decoding can be performed when the partition satisfies all the constraints by reading each row of the matrix;  $[x_{i,c}]_{N\times n}$  gives us the right colour for each number. Else,  $\boldsymbol{x}=[x_{i,c}]_{N\times n}$  will give potential partitions, for each possible resolution, each with some energy, and we can take the minimal configuration.

# 2 Hamiltonian Function 2

Alternatively, we can choose a different decision variable, where each variable can be thought of as a certificate.

**Definition 4.** The decision variable  $x_{m,k}$  is defined as

$$x_{m,n} := \begin{cases} 1 & Col(m) = Col(n) \\ 0 & else \end{cases}$$
 (8)

So,  $x_{i,j} = x_{j,i} = 1 \implies i$  and j have the same colour. Note that this does not impose any constraint on the number of colours necessary; only that in the most general sense, for a desired partition of N numbers,  $x_{i,j}$  have the same colour, or not.

Therefore, we can simplify the corresponding Hamiltonians as follows.

**Definition 5.** The Hamiltonian function for a Schur partition of the first N natural numbers is the following.

$$\mathcal{H}^C := A \sum_{i,j=1}^N x_{i,j} x_{i,i+j} x_{j,i+j}$$

**Definition 6.** The Hamiltonian function for a Weak Schur partition of the first N natural numbers is the following.

$$\mathcal{H}_{WS}^{C} := A \sum_{i,j=1}^{N} x_{i,j} x_{i,i+j} x_{j,i+j} \cdot (2N - |i-j|)$$

<sup>&</sup>lt;sup>1</sup>Usually local minima

In a similar manner to Eqn. 3, the |i-j| multiplier penalises the degenerate case of 2i = k, where the only possible way to decrease  $\mathcal{H}_{WS}^{C}$  is to maintain distinct colors for  $x_{i,j}$ ,  $x_{i,i+j}$  and  $x_{j,i+j}$ .

### 2.1 Decoding

Since the Hamiltonian does not impose the colour-constraint, the Decoding algorithm must. The proposal is a type of Branching algorithm, that has the signature: **Decode**:  $([x_{i,j}]_{N\times N}, n) \to C \in [N] \times [n]$ , where  $[N] = \{1, 2, \ldots, N\}$  and n is the number of colours.

Informally, the steps are as follows:

- 1. **Initialisation** Assign a random colour to the first number k = 1.
- 2. For each number k > 1, for each certificate  $x_{k,m} = 1$  where  $m \neq k$ ,
  - (a) if m has a colour, assign k the same colour and terminate.
- 3. If not terminated yet, for all certificates  $x_{k,m} = 0$   $(m \neq k)$ , create a list L = [1:n]
  - (a) for each  $x_{k,m} = 0$ , remove m from L.
  - (b) Branch on the size of L:
    - i. if |L| = 0, terminate with error.
    - ii. If |L| = 1 i.e.  $L = \{m\}$ , assign k the colour m and terminate.
    - iii. Else, branch on all possible values that remain in L.

### References

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