



HPC ISING MODEL SOLVER: APPLICATION IN MATERIAL DESIGN

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Introduction

The Density-Functional Theorem, used for modeling magnetic materials, usually provides us with a well-defined ground state of a given system, but not its transitional temperatures. For obtaining these properties, we present a generic and high-performance method for **mapping** an **ab-initio system** onto the **Heisenberg model** using our original software, JorGπ [1].

The most crucial part of the algorithm is identifying a set of **metastable states** of the system, achieved through solving a 3D Ising model. This being also the most computationally expensive part of the process, we **parallelized** it – using both distributed and shared memory models.

JorGπ: software for mapping DFT onto the Heisenberg model

Given the ground state of a magnetic structure, JorGπ:

1. processes an ab-initio model of the system,
2. generates a set of metastable states by solving the Ising model,
3. calculates the Heisenberg model from the output.

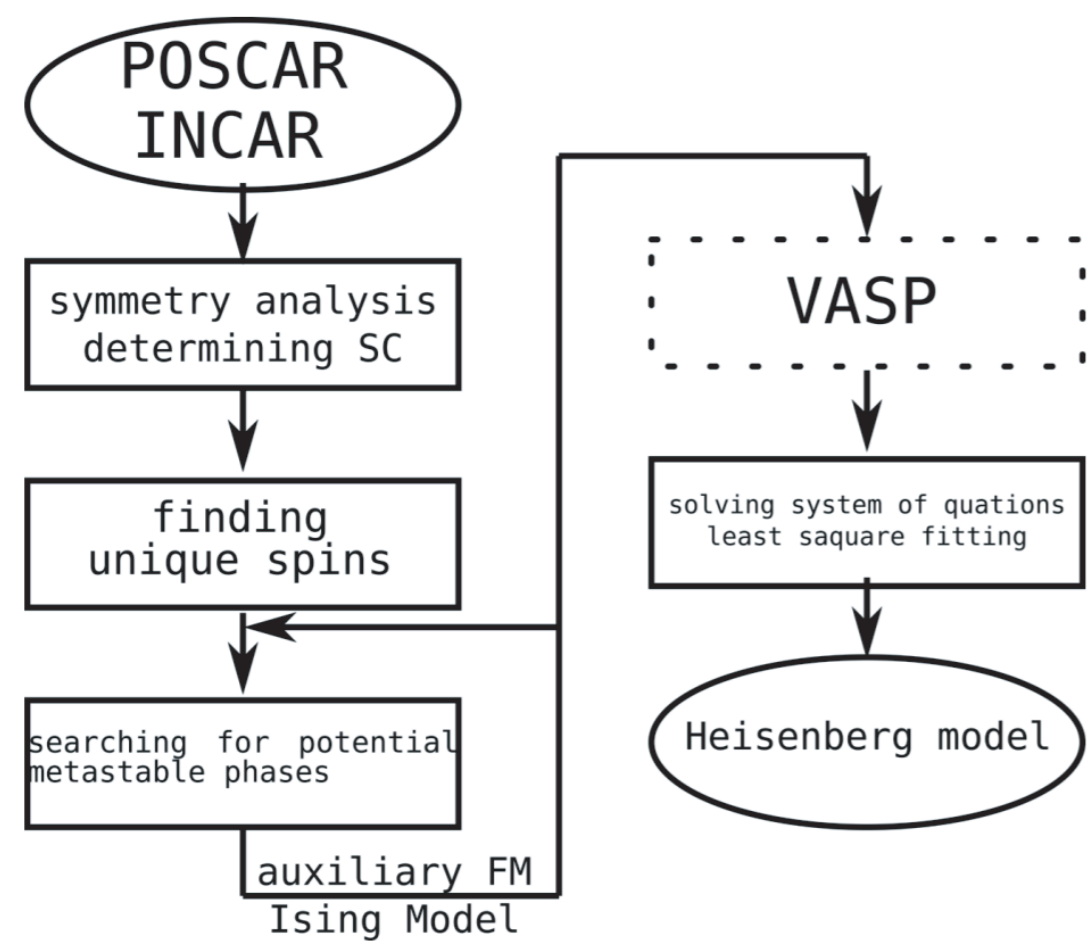


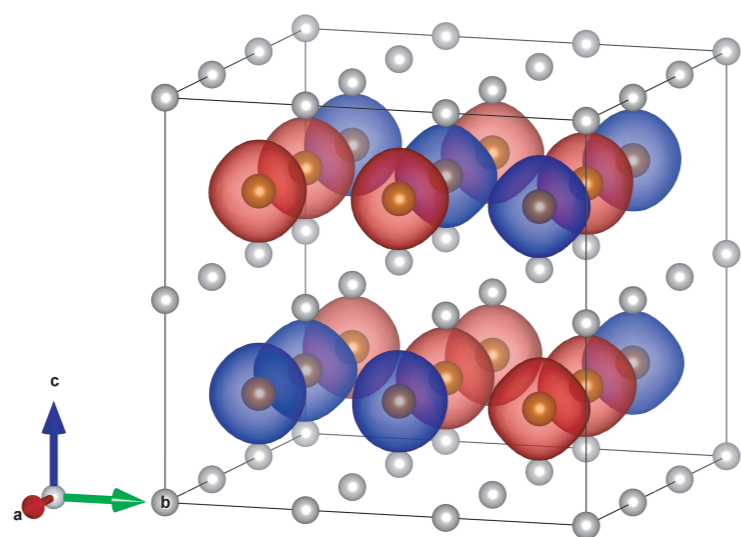
Fig. Workflow of JorGπ [1]

EXEMPLARY RESULTS - FEPT

Space group: P4/mmm (123):

a = b = 2.66084 Å
c = 3.68635 Å
α = β = γ = 90°

| | μFe(μB) | μPt(μB) |
|----------|---------|---------|
| FLEUR[6] | 2.866 | 0.384 |
| VASP[5] | 2.823 | 0.318 |
| JorGπ | 2.849 | 0.315 |



| | Distance(Å) | FLEUR[2] | SPR-KKR[3] | JorGπ | Ref. [4] |
|-------------|-------------|----------|------------|--------|----------|
| J[100](meV) | 2.661 | 25.73 | 31.52 | 35.77 | 22.25 |
| J[001](meV) | 3.686 | 8.82 | -3.72 | 13.28 | 1.15 |
| J[110](meV) | 3.763 | 19.89 | 17.56 | 19.24 | 18.58 |
| J[101](meV) | 4.546 | 8.20 | 8.58 | 15.67 | 16.03 |
| J[111](meV) | 5.268 | -15.32 | -10.58 | -21.07 | -19.75 |

JorGπ requires Python 3.6 with numpy 1.16.0, scipy 1.0.0, spglib 3.0.0, setuptools 40.8.0, and defusedxml. The Ising model solver is written in C++17 and utilizes GNU Scientific Library, OpenMP, and, optionally, MPI (the Message Passing Interface).



The Heisenberg model

Knowing the Heisenberg model, we can easily estimate e. g. T_C – the **Curie temperature** (above which a material loses its magnetic properties), hence, essentially, the temperatures at which the designed magnet will work.

The Heisenberg Hamiltonian is given by

$$H = -\frac{1}{2} \sum_{i \neq j} J_{ij} S_i S_j,$$

where J_{ij} are exchange interaction magnitudes of lattice site spins S_i .

- We can find the form of the Hamiltonian (i. e., its coupling parameters – J_{ij}) by identifying its metastable states.
- Furthermore, if $S_i S_j < 0 \Leftrightarrow J_{ij} < 0$, finding the excited states of the **Heisenberg model** is equivalent to finding the excited states of a **strictly ferromagnetic Ising model**.

Solving the Ising model

The **energy** of a spin configuration of the Ising model is described by the Hamiltonian

$$H = -\sum_{i \neq j} J_{ij} S_i S_j,$$

where J_{ij} are exchange interaction magnitudes of lattice site spins S_i .

Because the metastable states correspond to spin configurations with locally minimal energy, we can find them by minimizing the Hamiltonian.

To accomplish this, JorGπ performs **adaptive simulated annealing** (SA) [5] of a **ferromagnetic 3D Ising model**.

Parallel Solver

Since solving the Ising model is the most computationally expensive part of JorGπ (generating a single metastable state requires running SA at least once), we parallelized it.

- The parallel solver distributes the generation of states between **cluster nodes** via **MPI** (the Message Passing Interface). It employs a coarse-grained communication with SA running separately on each node.
- Additionally, **within** each **node**, loop calculations are parallelized via **OpenMP** using available **CPU cores**.
- To further increase efficiency, we intend to implement a hybrid method that combines parallel simulated annealing with a **genetic algorithm** [6] for improving initial states and, in this way, accelerating convergence.

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