

INS_IEP: A Matlab package for fitting Inelastic Neutron Scattering data.

Alban Bloor Riley¹, Marcus Webb¹, and Michael L. Baker²

¹ The University of Manchester, Department of Mathematics ² The University of Manchester, Department of Chemistry ¶ Corresponding author

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Summary

Inelastic neutron scattering (INS) is a spectroscopic technique used to measure the magnetic excitations in materials with interacting electron spins. Fitting the experimental data to a spin Hamiltonian model can be formulated as an inverse eigenvalue problem (IEP). INS_IEP is a MATLAB package that uses deflated numerical optimisation techniques to find multiple solutions to this problem. The package requires and is fully compatible with easyspin ([Stoll & Schweiger, 2006](#)), a package for solving similar problems in electron paramagnetic resonance.

Statement of need

Neutrons are an excellent bulk probe of material properties since they carry no charge and therefore penetrate deeply into matter. Neutrons also carry a quantum spin of a half, making them a sensitive probe of magnetism(?). Reactors and spallation sources with dedicated high-flux neutron sources serve the international research community with neutron scattering experiment capabilities for material research. One technique at neutron facilities is the measurement of inelastic neutron scattering (INS) to study magnetism. In an INS experiment, a sample under investigation is irradiated with a beam of neutrons and the scattered neutron energy and momentum transfer are detected. For samples composed of finite-size clusters of magnetic moment-carrying atoms, such as single ions or molecular-based magnets, the detected neutron energy transfer gives direct access to the quantum spin excitations ([M. Baker & Mutka, 2012](#); [Furrer & Waldmann, 2013](#); “[Spectroscopy Methods for Molecular Nanomagnets](#),” 2014). The energy of such excitations relates to the energy difference between eigenvalues of the Hamiltonian matrix that describes the quantum spin dynamics of the compound in question. Single-ion and molecular-based magnets are studied as prototype components (quantum bits, sensors) for quantum technologies. INS can therefore provide crucial information concerning the precise quantum properties of such systems. However, to relate the INS experimental results to the Hamiltonian that describes quantum spin dynamics requires parameterisation of matrix elements such that a set of eigenvalues and eigenstates matching the experiment are determined. This situation is known as the inverse eigenvalue problem. To date, this problem is addressed in an iterative process where parameters of the Hamiltonian are varied and the resultant eigenvalues compared with experiment. INS_IEP presents an elegant solution to solving this problem, alleviating the necessity of iterative Hamiltonian matrix diagonalisation, providing a more robust method to reliably extract an accurate spin Hamiltonian model from INS experimental data.

The Spin Hamiltonian

The Spin Hamiltonian, H , is an approximation of the Hamiltonian that uses spin coordinates instead of orbital coordinates, it is widely used to model data arising from many spectroscopy techniques (?). It can be modeled as a linear combination of interaction terms, we will focus on the zero field interaction, H_{ZFI} , and the electron-electron interaction, H_{EEI} :

$$H = H_{ZFI} + H_{EEI}$$

Both of these terms can themselves be modelled as the linear sum of other basis matrices. The zero field interaction can be written as:

$$H_{ZFI} = \sum_{-k \leq q \leq k} B_k^q O_k^q$$

where the O_k^q are Stevens Operators (Rudowicz & Chung, 2004), and B_k^q the associated parameter. When there are multiple spin centres it is necessary to take kronecker products of the operator with identity matrices of the appropriate for each other spin centre.

When there are multiple spin centres it is also necessary to include an electron-electron interaction term, $H_{\{EEI\}}$. This term will be the sum of interaction terms between each pair of spin centres:

$$H_{EEI} = - \sum_{i \neq j} J_{ij} S_i \cdot S_j$$

where S_i is the vector of spin operators $S_i = [S_x, S_y, S_z]$ for the i -th spin centre, and J_{ij} is the unknown parameter. Note that in the isotropic case J can be thought of as a scalar value, but in the anisotropic case will be a matrix where the off diagonals are skew symmetric. While the summation is in theory over all spin centre combinations, in practice many of these contributions will be negligible - often only the nearest neighbour interactions are modeled.

Mathematical Background

Inverse Eigenvalue Problem

The INS experiments provide eigenvalues of the Spin Hamiltonian matrix of the sample, the task of calculating the matrix from the eigenvalues is an inverse eigenvalue problem:

Let $A(x)$ be the affine family of matrices,

$$A(x) = A_0 + \sum_{i=1}^{\ell} x_i A_i,$$

where $x \in \mathbb{R}^{\ell}$ and $A_0, \dots, A_{\ell} \in \mathbb{R}^{n \times n}$ are linearly independent symmetric matrices, and denote the ordered eigenvalues of $A(x)$ as $\lambda_1(x) \leq \dots \leq \lambda_n(x)$. Then the least squares inverse eigenvalue problem is to find the parameters $x \in \mathbb{R}^{\ell}$ that minimises

$$F(x) = \frac{1}{2} \|r(x)\|_2^2 = \frac{1}{2} \sum_{i=1}^m (\lambda_i(x) - \lambda_i^*)^2$$

where $\lambda_1^* \leq \dots \leq \lambda_m^*$ are the experimental eigenvalues. In the case of INS fitting the A_i basis matrices will be a combination of Stevens operators and electron-electron exchange terms.

66 The IEP described above is formulated as an least squares problem, this is due to the fact that
67 the number of eigenvalues that can be probed by INS experiments is often a small subset of
68 the full spectrum

69 As far as we are aware this is the first time that the fitting of INS data has been explicitly
70 formulated as an IEP. The advantage of this formulation is that there are explicit formulas for
71 the derivatives of $r(x)$, the first derivative (Jacobian) is:

$$J_r(x) = \begin{pmatrix} q_1(x)^T A_1 q_1(x) & \dots & q_1(x)^T A_\ell q_1(x) \\ \vdots & \ddots & \vdots \\ q_m(x)^T A_1 q_m(x) & \dots & q_m(x)^T A_\ell q_m(x) \end{pmatrix}.$$

72 And the second derivative (Hessian) is:

$$(H_r)_{ij} = 2 \sum_{k=1}^m (\lambda_k - \lambda_k^*) \sum_{\substack{t=1 \\ \lambda_t \neq \lambda_k}}^m \frac{(q_t^T A_i q_k)(q_t^T A_j q_k)}{\lambda_k - \lambda_t}.$$

73 Access to analytical forms of the derivatives means it is not necessary to use the finite difference
74 approximation that other approaches use, making the three optimisation methods that INS_IEP
75 uses faster and more accurate.

76 Methods

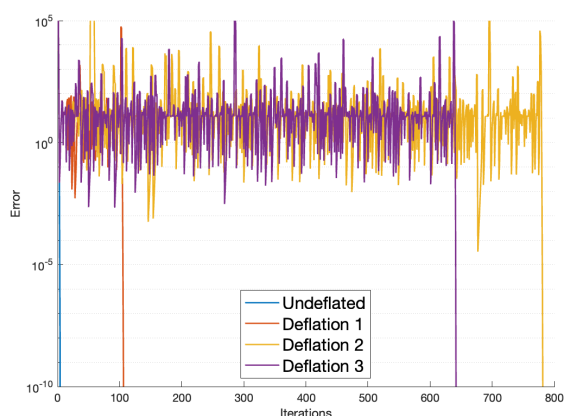
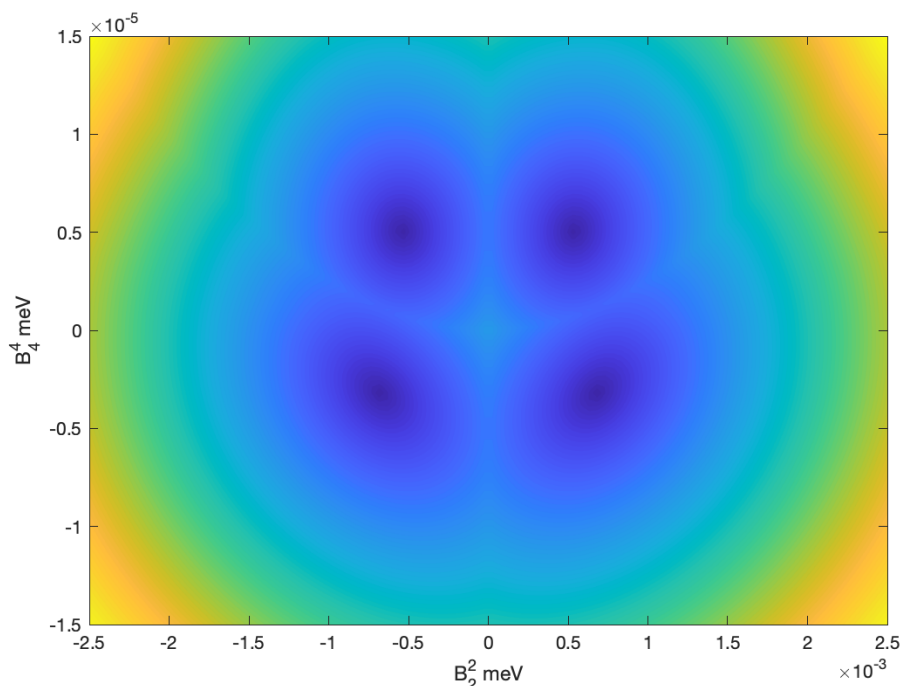
77 All of the methods used are iterative schemes of the form $x^{k+1} = x^k + p^k$ where the step p^k
78 uniquely defines each algorithm:

- 79 ■ Newton's method: $p^k = (J_r^T J_r + H_r r) - 1 J_r^T r$ (Nocedal & Wright, 2006)
- 80 ■ Gauss-Newton method: $p^k = (J_r^T J_r)^{-1} J_r^T r$ (?)
- 81 ■ Lift and Projection Method: $p^k = B^{-1} J_r^T r$ (?)

82 Where the matrix B is the Gram matrix formed from the frobenius inner products of the
83 basis matrices: $B_{ij} = \langle A_i, A_j \rangle_F$. The Lift and Projection method is a Riemannian Gradient
84 descent method (?), inspired by the Lift and Projection method (Chen & Chu, 1996), specifically
85 designed for solving IEPs. In (?) it is proven that the method is a strictly descending algorithm,
86 that is it reduces the value of the objective function every step. Both the deflated Gauss-Newton
87 method and the Riemannian Gradient descent Lift and Projection method are new methods
88 designed for this package.

89 Deflation

90 The number of eigenvalues that can be probed via INS experiments varies depending on the
91 equipment and sample in question, meaning that the fitting problem is often under (or even
92 over) determined. The IEP is also highly nonlinear and due to the experimental nature of
93 the data there is no guarantee that the problem is not ill-posed. One consequence of this is
94 that the solution space may be very 'bumpy', that is there may exist many local minimisers
95 to the problem. For example in figure ??, there are clearly 4 distinct solutions (for more
96 details see Example1_Mn12 in the examples folder). We seek to solve the problem of multiple
97 local minima by the use of Deflation, a numerical technique used to find multiple solutions to
98 systems of equations (P. E. Farrell et al., 2015; Patrick E. Farrell et al., 2020). Fortunately it
99 is cheap to apply deflation for the above methods, it is simply a change to the length of the
100 step - notably this means that the direction of each step does not change (?).



Examples

Example 1 - Mn12

The first example we will look at is Manganese-12-acetate. This is a well known example in the INS and magnetism community, as one of the first molecules that behaves like a nano-sized magnet with a molecular magnetic coercivity as well as its role in the research of quantum tunnelling of magnetisation (Friedman et al., 1996; Sessoli et al., 1993).

The Spin Hamiltonian of this system, using the giant spin approximation, can be represented as a 21×21 matrix modelled using 4 Stevens operators (Bircher et al., 2004):

$$H = B_2^0 O_2^0 + B_4^0 O_4^0 + B_2^2 O_2^2 + B_4^4 O_4^4 \in \mathbb{R}^{21 \times 21}$$

We utilise the same spin system syntax as easyspin, so to set up the problem we first set up the model, along with initial guesses for the parameters:

```
%One spin centre (because giant spin approximation)
```

```
Sys0.S=10;
```

```
%Four Stevens operators
```

```
Sys0.B2 = [-100,0,-1000,0,0];
```

```
Sys0.B4 = [-1,0,0,0,-1,0,0,0,0];
```

113 Then we input the experimental eigenvalues, these will normally be translated such that the
114 smallest eigenvalue is zero, and define which parameters to fit. Note that all values given must
115 be in Hertz, so it may be useful to use conversions.

```
rcm = 29979.2458; meV = rcm*8.065; %Conversions values
```

```
%Input calculated eigenvalues:
```

```
Exp.ev = [0,0,1.24,1.24,2.3,2.3,3.18,3.18,3.91,3.91,4.5,4.5,4.97,4.97,5.32,5.32,5.54,5.54];
```

```
%Vary all non zero parameters (no Fixed parameters):
```

```
Vary = Sys0;
```

116 Then all that is required is to call INS_IIEP with these three inputs:

```
SysOut = INS_IIEP(Sys0,Vary,Exp);
```

117 If we wish to find all four solutions as shown in figure ?? then we use the additional option:

```
Opt.NDeflations = 4;
```

```
SysOut = INS_IIEP(Sys0,Vary,Exp,Opt);
```

118 In this case SysOut will be an array of four spin structures each containing a distinct solution.

119 Example 2 - Chromium(iii) Horseshoes

120 The second example concerns antiferro-magnetically coupled chromium(III) chains six atoms
121 long (M. L. Baker et al., 2011), although different length chains are of interest and can also
122 be modelled. Because there are multiple spin centres an electron-electron interaction term is
123 required. The spin hamiltonian is a 4096×4096 matrix composed of two Stevens operators
124 and one interaction term:

```
Sys0.S = [1.5 1.5 1.5 1.5 1.5 1.5];
```

```
Sys0.B2 = [1 0 -1 0 0;  
1 0 -1 0 0;  
1 0 -1 0 0;  
1 0 -1 0 0;  
1 0 -1 0 0;  
1 0 -1 0 0];
```

```
Sys0.J = [100,0,0,0,0,100,0,0,0,100,0,0,100,0,100];
```

```
Vary = Sys0;
```

```
Exp.ev = [0,0.355,0.457,0.497,1.576,1.577,1.592,1.629,1.632,2.97,2.98,3.002,3.004,  
3.01,3.038,3.821,3.824,3.827,3.837,3.856,3.879,3.888,3.895,3.903];
```

125 Note that only 24 eigenvalues were found experimentally, so this will form a partial eigenvalue
126 problem. To find the solution system is as simple as:

```
SysOut = INS_IIEP(Sys0,Vary,Exp);
```

127 It is possible to find multiple minimising systems even if they do not make any sense physically,
128 however due to the scaling of the problem a change in the default deflation parameters is
129 necessary:

```
Opt = struct('NDeflations',4,'Sigma',1e-7);
```

```
SysOut= INS_IIEP(Sys0,Vary,Exp,Opt);
```

130 The output contains four different spin systems that all have the same eigenvalues as input,
131 one is the original solution up to a change of sign of the B_2^2 parameter, and all have the same

exchange term.

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