

Chemical design of molecular nanomagnets using data mining, dashboards, and statistics as a strong framework

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Abstract

After three decades of continuous study, the magnetic memory in molecules has been cooled to the temperature of liquid nitrogen. A smart selection of the magnetic ion and the coordination environment is necessary to increase this operating temperature. But the biggest contributors have been chance, oversimplified ideas, and chemical intuition. We gathered chemical and physical data for lanthanide-based nanomagnets to create a catalogue of over 1400 published experiments, developed an interactive dashboard (SIMDAVIS) to visualize the dataset, and applied inferential statistical analysis to it in order to establish a strong framework for statistically driven chemical design. We discovered that the magnetic memory exhibits a strong link with the effective energy barrier calculated from the Arrhenius equation, and that among all chemical families

Keywords: molecular nanomagnets, datamining

Introduction

With the goal of facilitating the chemical design of lanthanide single-ion magnets (SIMs), herein we mined experimental data from over 450 articles from the first 17 years of SIM research and applied data visualization tools followed by state-of-the-art statistical analysis. This allowed us to obtain robust statistically-driven insights on the key factors governing the slow relaxation of the magnetization.

Additionally, we created a dataset and a dashboard: a free, openly accessible online interactive app that allows any user to perform qualitative analyses as well as to browse the dataset. This statistical study represents a quantitative and data-supported verification/refutation of several widely held "chemical intuitions" in the field of molecular magnetism.

A brief history of SIMs

Molecular nanomagnets were reported for the first time at the beginning of the 1990s, when $Mn_{12}O_{12}(CH_3COO)_{16}(H_2O)_4$, a polynuclear magnetic complex with strong magnetic coupling between d-block ions was reported to display magnetic hysteresis, similar to classical magnets, but with a quantum tunnelling mechanism for the relaxation of the magnetisation.^{1,2} This was the first of a plethora of single-molecule magnets

(SMMs), the term coined for systems behaving as hard bulk magnets below a certain temperature, but where the slow relaxation of the magnetization is of purely unimolecular origin. The magnetic

behavior in SMMs can be approximated to that of an effective anisotropic magnetic moment arising from the exchange interactions between the spins of the metal ions. The reversal of this giant anisotropic spin occurs by populating excited spin states and overcoming an energy barrier. Hence, the thermal dependence of the relaxation rate was described by the Arrhenius equation (see Fig. 1), using this effective energy barrier (U_{eff}) and a pre-exponential factor (τ_0).³ Both parameters

were not extracted directly from the hysteresis loop, but rather from the combined frequency- and temperature-dependence of the so-1).^{3,4} The experimental fact that the magnetic hysteresis was observed only near liquid-helium temperatures was rationalized mainly by the typically low values found for the effective energy barrier $U_{eff} \approx 50$ K.⁵ Initial models based on effective spin Hamiltonians gave rise to the relation $U_{eff} = DS_z^2$ and concluded that the best strategy to raise U_{eff} and, therefore, to improve the maximum hysteresis temperature (T_{hmax}) is to maximize the total effective spin (S), rather than the magnetic anisotropy (D). Indeed, the latter is a less straightforward

target for the synthetic chemist.⁶ Despite great effort toward the synthesis of such systems and an abundance of molecules with increasing values of S , very little progress was made in the first decade in terms of increasing U_{eff} for T_{hmax} .⁷

In the 2000s, a second generation of molecular nanomagnets emerged, namely bis-phthalocyanines (Pc) "double deckers".⁸ This novel type of SMMs, commonly known as Single Ion Magnets (SIMs), is based on mononuclear complexes containing a single magnetic ion embedded in a coordination environment, thus constituting the smallest molecular nanomagnet. Their properties arise from a strong spin-

orbit coupling which, combined with the crystal-field interaction with the surrounding ligands, results in an enhanced magnetic anisotropy when compared to the first generation of SMMs. Identical data treatment using the Arrhenius equation resulted in effective energy barriers U_{eff} up to an order of magnitude higher for SIMs based on rare-earth ions when compared to those of polynuclear metal complexes of the f -block.

Similarly, the characteristic maxima in the out-of-phase component of the ac susceptibility χ'' moved to higher temperatures, albeit the T_{hits} did not increase significantly.

After the germinal LnPc_2 , different chemical families such as polyoxometalates⁹ and metallocenes¹⁰ were synthesized and were shown to exhibit slow relaxation of the magnetization of purely molecular origin (Fig. 2). The realization that lanthanide SIMs were not restricted to a single chemical strategy inspired a large community of chemists. As a result, in the last 15 years, SIM behavior was reported in over 600 compounds, whilst above a third of these compounds displayed magnetic hysteresis. Recent efforts have been made to offer some perspective,¹¹⁻¹⁴ but so many studies pursuing independent inspirations have been reported that it is hard for any single scientist to distinguish anecdotal claims from proven strategies. Modern techniques of data analysis and visualization can contribute to remedy this knowledge gap. In particular, dashboards are intuitive graphical software applications for dynamic data visualization and information management, of growing popularity in different fields.¹⁵⁻¹⁷

The present work aims to rationalize the correlations among the different physical variables involved in SIMs. A common working hypothesis is that the parameters arising from the ac magnetometry (e.g., U_{eff}) are well correlated with the experimental values (e.g., T_{hits}). This, however, has not been proven and has actually been challenged in various ways.^{14,28,29} Over the years, various theoretical approaches have put the focus on the role of different physical processes and different parameters.²⁸ In order to provide the synthetic chemists with a statistically-driven chemical design guide, we applied the techniques of third-generation computational chemistry,³² starting by collecting a high-quality dataset and representing the data in an interactive dashboard.

Statistically-driven chemical design of SIMs

SIMDAVI¹ allows the visualization of the relationships between chemical and physical variables

in SIMs, and thereby enables determining the main variables that the synthetic chemist needs to consider to obtain the desired physical properties. We will first analyze this qualitatively employing a series of boxplots, violin plots and bar charts (see Fig. 4 and Supplementary Figs. 11.1-11.6, 12). The full statistical analysis is presented in Supplementary Sections 4, 5 and 6.

First, let us focus on the effective energy barrier U_{eff} and the blocking temperature T_{B3} (the temperature for maximum out-of-phase ac susceptibility χ'' at 10^3 Hz, see Fig. 1). From Fig. 4a and Supplementary Figs. 11.1-11.4, we can see that the only chemical family with a clearly distinct behavior is the LnPc_2 family, with median values of $U_{\text{eff}} > 300$ K and $T_{\text{B3}} > 30$ K. Equivalently, one can see that Dy^{3+} and Tb^{3+} are somewhat better than the rest, and that in general oblate ions perform better than prolate ions, for both properties. In addition, non-Kramer's ions present higher median T_{B3} but similar U_{eff} values compared with Kramer's ions.

Now, let us analyse the maximum hysteresis temperature T_{hits} which has been much less studied despite being the main justification for this whole field. The only chemical family with a distinct positive behavior is the metallocene family. More surprisingly, Er^{3+} complexes have distinctly high hysteresis temperatures, markedly with a higher median than Dy^{3+} or Tb^{3+} complexes. This is in sharp contrast with their relative T_{B3} values which are consistently much lower in the case of Er^{3+} complexes. This not only indicates that searching for equatorial environments, precisely the ones that favor good magnetic properties in Er^{3+} complexes,³⁰ often results in more rigid ligands, but also indicates an underexplored territory. It is certainly possible that chemical modifications of $[\text{Er}^{(COT)}_2]$ (or other Er^{3+} record-bearing

complexes) designed to optimize the detrimental effect of molecular vibrations may achieve records that are competitive with DyCp_2 . Prolate ions are consistently and surprisingly better than the oblate ones, having a higher median value for T_{hits} . This is again in contrast with the opposite behavior which is observed for T_{B3} and U_{eff} , and possibly again due to the influence of Er^{3+} complexes with the more rigid equatorial environments. Finally, the coordination number and the number of ligands does have an influence on the statistically expected hysteresis temperature, with the best ones being 2 and 7 in the case of the coordination number and just 7 for the number of ligands. As we will discuss below, there are chemical insights to be gained from this.

To put all these trends into perspective, it is important to numerically analyse the connection between the different variables and the clustering of four data. A lognormal analysis (see Supplementary Section 4.3) shows that the three main chemical variables, namely the chemical family, the lanthanide ion and the coordination elements, are sufficient to reasonably explain the variation of values of the others, meaning that ϵ is a limit on the information one can independently extract from the rest of the chemical variables. Multiple correspondence analysis (see Supplementary Sections 4.1, 4.2) suggests a chemical clustering that consists in three small groups, namely Gd^{3+} complexes, metallocene's and $LnPc_2$ double deckers, and two much larger groups with a large overlap with oblate and prolate ions respectively. A factorial analysis of mixed data considering also all magnetic information available (see Supplementary Section 6) simplifies the clustering to three groups. Again, the two distinct families present a large overlap with metallocene's and $LnPc_2$ double decker chemical families, both of them presenting significantly better properties than the other kinds of samples.

Further insight is provided by bar charts representing the reported presence of magnetic hysteresis, whether full or pinched, as a function of different chemical variables (Fig. 4). Note that we are limited by the minority of the samples where hysteresis or its absence is reported; in the vast majority of the cases this information is lacking. Nevertheless, here it is apparent that certain families such as $LnPc_2$ (and metallocene's) tend to display (pinched) hysteresis.

The effective energy barrier: oversimplified yet meaningful

A key question is how much the analyses in this field have been affected by the simplified assumption that SIMs relax via an Orbach mechanism, which is characterized by τ_0 and U_{eff} . It has been pointed out that frequently, as U_{eff} increases, τ_0 decreases, leaving relaxation times essentially constant.⁵ The classical text of Abraham and Blaney offered the following relation between the two parameters for the two-phonon Orbach process:³⁴

$$\frac{1}{\tau_0} = R_{Or} \cdot (U_{eff})^n \quad (1)$$

where $n = 3$, reasonable parameters for rare earth elements resulted in an Orbach rate $R_{Or} \approx 10^4 \text{ K}^{-3 \cdot \text{s}^{-1}}$, and early experimental results were in the range $10^3 \text{ K}^{-3 \cdot \text{s}^{-1}} < R_{Or} < 10^5 \text{ K}^{-3 \cdot \text{s}^{-1}}$. Fitting τ_0 vs U_{eff} in our dataset to equation (1) results in $n \approx 2.4$, $R_{Or} \approx 10^3 \text{ K}^{-3 \cdot \text{s}^{-1}}$

(see Supplementary Table 5). This minor discrepancy with the expected exponent serves as an independent evaluation of the limitations of a simple Orbach model. We also find $R_{Or}(\text{prolate}) \approx 5 \cdot R_{Or}(\text{oblate})$, meaning that, for comparable U_{eff} , relaxation for oblate ions is on average substantially slower than that for prolate ions. This is consistent with the observation that complexes of oblate ions present values of T_{B3} higher than expected considering their U_{eff} (see Supplementary Fig. 11.3). The limited (<100) data points of U_{eff}, ff, τ_0, ff pairs, where all relaxation processes were considered, present a better agreement on the exponent, with $n \approx 3$ and lower Orbach rates $R_{Or} \approx 150 \text{ K}^{-3 \cdot \text{s}^{-1}}$. The remaining crucial issue is to quantify up to what level the value of U_{eff} and τ_0 are well correlated with the slow relaxation of the magnetisation, or to determine whether one would need to employ $U_{eff, ff}$ instead. Let us proceed with increasing the order of complexity. A visual inspection in SIMDAVIS shows that, in a few cases where there is simultaneous information on U_{eff} and $U_{eff, ff}$, their values are very similar (Fig. 5a). Furthermore, this partial information is corroborated by the very similar dependencies of T_{B3} or T_{hyst} vs either U_{eff} or $U_{eff, ff}$, as well as in the numerical correlations (see Supplementary Sections 3.2 and 5.3). A categorical analysis (Figs. 5b, c) shows that the data dispersion is large, meaning that it is impossible to predict the experimental behaviour for a individual sample merely from its U_{eff} value. However, it demonstrates that, statistically, samples which present a maximum in the out-of-phase susceptibility χ'' , or hysteresis, also present higher U_{eff} values. A more thorough numerical analysis (see Supplementary Section 6) confirms these trends.

An in-depth statistical analysis of all physical parameters based on the Akaike Information Criterion (see Supplementary Section 5.3) concludes that U_{eff} derived from a simple Arrhenius plot is the best single predictor for the magnetic behaviour in our dataset. This means that, whether we are discussing in terms of the out-of-phase component of the ac susceptibility or magnetic hysteresis, U_{eff} is a better predictor than τ_0 , τ_0 , ff , $U_{eff, 2}$ and, in practice, than $U_{eff, ff}$. Factorial analysis of mixed data (see Supplementary Section 6) also reveals the predictive power of U_{eff} compared with τ_0 . Note that this does not contradict previous studies which demonstrated that a variation in the Orbach barrier does not fully explain the differences in retention of magnetisation,¹⁴ since we have not explicitly considered other relaxation mechanisms in the present work. temperatures. It is possible that the next family of record-

setters is related to one of the promising candidates in Fig. 6. Two axial phosphine oxide ligands with bulky substituents seem to function in a similar way as metallocene's, despite the five equatorial IH_2O molecules.^{24,37} This strategy is not restricted to phosphine oxides and deserves to be explored further: as can be seen in Supplementary Fig. 11.4, complexes with 7 ligands have median values of T_{hyst} close to 10 K, as high as those with 2 ligands.

At the same time, here we provide a catalogue of lanthanide SIMs, together with SIMDAVIS, a dashboard that allows its interactive navigation; this is a type of tool utterly missing in the field of molecular nanomagnets. Perhaps more importantly in the wider perspective of design of new materials³⁸ and new molecules,^{32,41} the dataset curated in this work will serve for Machine Learning studies and can also be employed as an annotated training data set for the development of new web scraping systems to retrieve chemical data,^{42,43} or even word embeddings⁴⁴ from the scientific literature. Finally, this work constitutes a step towards the availability of findable, accessible, interoperable, and reusable (FAIR) data in Chemistry.⁴⁵

METHODS

Data gathering.

This process started with the collection and organisation of literature data. The following search criterion was applied for the manuscript: articles are searched via Web of Science, employing this code:

TOPIC: TS=((lanthan* OR 4\$f OR "rare\$earth") AND ((single NEAR/1 magnet*) OR "slow relaxation"))
Timespan:2003-2019

For an article to be included in the study, it needs to contain data on at least one compound with certain requirements as follows: (a) contain one trivalent lanthanide ion from the set $\text{Ln}=\{\text{Pr}, \text{Nd}, \text{Sm}, \text{Gd}, \text{Tb}, \text{Dy}, \text{Ho}, \text{Er}, \text{Tm}, \text{Yb}\}$ and (b) contain no other paramagnetic entity with the only accepted exception being the presence of a single radical in the coordination sphere and (c) present no strong $\text{Ln}-\text{Ln}$ interaction, in particular meaning the $\text{Ln}-\text{Ln}$ distance needs to be larger than 5 Å and more than 3 bridging atoms between neighbouring Ln centres, and there cannot be a radical in the bridge. Additionally, the data need to include at least one of the following information: (a) whether χ'' presents a maximum as a function of T , or a mere frequency-dependence, or neither; (b) χ'' vs T with at least one frequency (f) in the window $0.9\text{kHz} \leq f \leq 1.1\text{kHz}$ and at a field (B) in the window $0 \leq B \leq 2\text{ T}$;

(c) U_{eff} ; (d) the presence or absence of hysteresis; (e) T_{hyst} at sweep speeds (v) in the window $0.05\text{ T/s} \leq v \leq 0.3\text{ T/s}$. The compounds were classified in chemical families: LnPc_2 , polyoxometalates, Schiff base, metallocenes, diketonates, radicals, TM near Ln , mixed ligands, and other families. Furthermore, we registered for each sample (when available), the lanthanide ion, its concentration, the coordination number and number of ligands coordinated to the lanthanide ions, the coordination elements, the presence of a field-dependent χ'' or a maximum, the temperature of said maximum in presence or absence of an external magnetic field, the external magnetic field, the extracted effective energy barrier and relaxation time, either from a simplified Arrhenius fit or from a model considering all relaxation processes, whether these were extracted from the maxima of χ'' vs T at different frequencies or from an Argand fit, the presence of hysteresis in the magnetisation, and the maximum temperature at which it was recorded. Additionally the DOI, the full reference to the original article, and a link to a CIF file were recorded for each sample. Further details including the classification in chemical families and the criteria for data extraction are provided in Supplementary Sections 1 and 2.

SIMDAVIS dashboard. We programmed the dashboard using R language^{51,52} and shiny,⁵³ an open-source R package to create the interactive web app. The design aimed to obtain a clean and simple user interface that adapts automatically to any screen size. The R packages read,⁵⁴ dplyr,⁵⁵ DT,⁵⁶ ggplot2⁵⁷ and rcrossref⁵⁸ were also employed in the development of the dataset or the app. The dashboard-style web application is available at <https://go.uv.es/rosaleny/SIMDAVIS>. This interface allows for variables in the analysis, and subsets of the data, to be adjusted and chosen in real time.

Statistical analysis. The statistical analysis was also based on R, a widely used software environment for statistical computing and graphics, and included the Gifi system for Multiple Correspondence Analysis⁵⁹ (R homals package,⁶⁰ ade4 package,⁶¹ see details in Supplementary Section 4.1), hierarchical clustering studies (FactoMineR,⁶² see details in Supplementary Section 4.2), lognormal modelling (Poisson's distribution, see Supplementary Section 4.3), factorial analysis of mixed data (FactoMineR and factoextra,⁶³ see details in Supplementary Section 6) as well as Pearson's product-moment correlation and the Akaike information criterion (AIC)⁶⁴ (see details in Supplementary Section 5.3). The analysis was repeated and verified an overall excellent qualitative and quantitative consistency in all

results between the period 2003-2017 (1044 samples) and 2003-2019 (1405 samples).

Program of Excellence and CDEIGENT/2019/022). The statistical analysis was performed by RaquelGavidiaJosawiththeStatisticalSectionoftheS. C.S.I.E.(UniversitatdeValència).TOCfigurecreated withBioRender.com.

Results

An interactive dashboard for lanthanide SIMs

In this work, we built a dataset of the most relevant chemical and physical properties of 1405 lanthanide SIM samples collected from 451 scientific articles (Supplementary Section 7) published between 2003 and 2019 and developed a user-friendly dashboard-style web application named SIMDAVIS (Single Ion Magnet DAta VISualization) to host it. The dataset contains over 10000 independent pieces of chemical information, as well as over 5000 independent pieces of physical (magnetic) information. Furthermore, the dataset is hierarchically clustered into magnetostructural “taxonomies” (see Supplementary Sections 4 and 6) in order to pave the way for further analysis, including Machine Learning studies. SIMDAVIS allows the chemical community to visualize the key relationships between chemical structures and physical properties in our catalogue of SIMs. Our interactive dashboard can be directly invoked by accessing the internet site where it is located.³³ It is organized in 6 main tabs: Home, ScatterPlots, BoxPlots, BarCharts, Data (View Data and Download Data) and About SIMDAVIS (Variables, Authors, Feedback&Bugs, Changelog and License) as we can observe in Fig. 3. In the SIMDAVIS dashboard, the most versatile source of graphical information is the “ScatterPlots” tab,. The next two tabs display the data in complementary ways. The “BoxPlots” tab allows to examine the distribution of each SIMs quantitative property vs a categorization criterion, e.g. we can see the distribution of U_{eff} values as a function of the coordination elements. The boxplot for each category is shown, including the median and the interquartile range. The “BarCharts” tab allows the exploration of the frequency of different qualitative variables in our dataset. Stacked bar graphs allow the simultaneous analysis of two qualitative variables, e.g. we can display, for each chemical family, the number of samples which present magnetic hysteresis. The “Data” tab is a powerful interface to browse the dataset, featuring the possibility to choose the data columns to show, ordering in ascending or descending order, and filtering by arbitrary keywords; it also permits downloading all data, including links to the CIF

files, when available. Finally, the “About SIMDAVIS” tab contains information about the variables contained in the dataset.

Conclusions

We have systematically analysed over 450 articles to collect information from over 1400 samples reported over the first 17 years of the field of lanthanide-based SIMs and built a user-friendly dashboard for the visualization of all the collected data. Moreover, we carried out an in-depth statistical analysis that allowed extracting trends, distinguishing the most relevant variables and grouping the data in clusters based on their chemical and physical properties. From this study, we can highlight two main pieces of information. In the first place, from the point of view of the parametric characterisation, the simple Arrhenius fit assuming an Orbach process has been proven to be surprisingly meaningful, with the expected relation between τ_0 and U_{eff} . One can therefore perform this oversimplified theoretical fit knowing that the effective energy barrier U_{eff} has been proven to present a consistently good correlation with SMM behaviour. Crucially, we have also shown the very different nature of short-term magnetic memory in form of the blocking temperature T_{B3} at 10 3 Hz and its long-term counterpart in the form of maximum hysteresis temperature T_{Hyst} . The best strategies that optimize the former are not necessarily the best for the latter. In the second place, the chemical roadmap for the preparation of lanthanide complexes with higher T_{Hyst} becomes now a little clearer. There is a single chemical strategy that consistently produces good magnetic memories, namely sandwiching an oblate ion between two rigid, planar, aromatic ligands; furthermore, the ion should be chosen to result in the most favourable MJ structure, given the electron distribution offered by the ligand. Up to now, only two chemical families are well adapted to this strategy, namely TbPc2 complexes and dysprosium metallocenes.

Optimization is ongoing within these two families, for example TbPc2 complexes featuring a radical Pc display enhanced properties,³⁵ and the reduced (divalent) analogues of DyCp2.³⁶ We find comparatively little value in further pursuing chemical strategies that have been amply explored and never yielded hysteresis above 10 K. On the other hand, we also evidence that there is, of course, value in chemical ingenuity and exploration, in the quest for another successful strategy, which according to our results might well be based on equatorial erbium complexes, since these display consistently high T_{Hyst} values. Note that a few complexes included in our data fall into

ample families such as “mixed ligands” or “other families”, and yet present excellent hysteresis Page 8/19 temperatures. It is possible that the next family of record-setters is related to one of the promising candidates in Fig. 6. Two axial phosphine oxide ligands with bulky substituents seem to function in a similar way as metallocenes, despite the five equatorial H₂O molecules. 24,37 This strategy is not restricted to phosphine oxides and deserves to be explored further: as can be seen in Supplementary Fig. 11.4, complexes with 7 ligands have median values of Thyst close to 10 K, as high as those with 2 ligands. At the same time, here we provide a catalogue of lanthanide SIMs, together with SIMDAVIS, a dashboard that allows its interactive navigation; this is a type of tool utterly missing in the field of molecular nanomagnets. Perhaps more importantly in the wider perspective of design of new materials 38–40 and new molecules, 32,41 the dataset curated in this work will serve for Machine Learning studies and can also be employed as an annotated training data set for the development of new web scraping systems to retrieve chemical data, 42,43 or even word embeddings, 44 from the scientific literature. Finally, this work constitutes a step towards the availability of findable, accessible, interoperable, and reusable (FAIR) data in Chemistry. 45

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