




# 1 SpiDy.jl: open-source Julia package for the study of 2 non-Markovian stochastic dynamics

3 **Stefano Scali** <sup>1</sup>¶, **Simon Horsley** <sup>1</sup>, **Janet Anders** <sup>1,2</sup>, and **Federico  
4 Cerisola** <sup>1,3</sup>

5 **1** Department of Physics and Astronomy, University of Exeter, Exeter EX4 4QL, United Kingdom **2**  
6 Institute of Physics and Astronomy, University of Potsdam, 14476 Potsdam, Germany **3** Department of  
7 Engineering Science, University of Oxford, Oxford OX1 3PJ, United Kingdom. ¶ Corresponding author

DOI: [10.xxxxxx/draft](https://doi.org/10.xxxxxx/draft)

## Software

- [Review](#) 
- [Repository](#) 
- [Archive](#) 

Editor: [Matthew Feickert](#) 

## Reviewers:

- [@Datseris](#)
- [@mdavezac](#)
- [@dawbarton](#)

Submitted: 05 October 2023

Published: unpublished

## License

Authors of papers retain copyright  
and release the work under a  
Creative Commons Attribution 4.0  
International License ([CC BY 4.0](#)).

## 8 Summary

9 SpiDy.jl solves the non-Markovian stochastic dynamics of interacting classical spin vectors  
10 and harmonic oscillator networks in contact with a dissipative environment. The methods  
11 implemented allow the user to include arbitrary memory effects and colored quantum noise  
12 spectra. In this way, SpiDy.jl provides key tools for the simulation of classical and quantum open  
13 systems including non-Markovian effects and arbitrarily strong coupling to the environment.  
14 Among the wide range of applications, some examples range from atomistic spin dynamics to  
15 ultrafast magnetism and the study of anisotropic materials. We provide the user with Julia  
16 notebooks to guide them through the various mathematical methods and help them quickly  
17 set up complex simulations.

## 18 Statement of need

The problem of simulating the dynamics of interacting rotating bodies and harmonic oscillator  
networks in the presence of a dissipative environment can find a vast range of applications  
in the modeling of physical systems. This task is rendered particularly challenging when one  
desires to capture the non-Markovian effects that arise in the dynamics due to strong coupling  
with the environment. SpiDy.jl is a library that allows the user to efficiently simulate these  
systems to obtain both detailed dynamics and steady-state properties.

A relevant example of the applicability of SpiDy.jl is the modeling of spins at low temperatures  
and at short timescales, which is a fundamental task to address many open questions in the  
field of magnetism and magnetic material modeling ([Halilov et al., 1998](#)). State-of-the-art  
tools such as those developed for atomistic spin dynamics simulations are based on solving the  
Landau–Lifshitz–Gilbert (LLG) equation ([Evans et al., 2014](#)). Despite their massive success,  
these tools run into shortcomings in accurately modeling systems at low temperatures and  
for short timescales where environment memory effects have been observed ([Ciornei et al.,  
2011](#); [Neeraj et al., 2020](#)). Recent work has focused on developing a comprehensive quantum-  
thermodynamically consistent framework suitable to model the dynamics of spins in magnetic  
materials while addressing these shortcomings ([Anders et al., 2022](#)). This framework includes  
strong coupling effects to the environment such as non-Markovian memory, colored noise,  
and quantum-like fluctuations. At its core, SpiDy.jl implements the theoretical framework  
introduced in ([Anders et al., 2022](#)), allowing for the study of environment memory effects  
and anisotropic system-environment coupling. SpiDy.jl can be readily adopted for atomistic  
spin dynamics simulations ([Barker & Bauer, 2019](#); [Evans et al., 2014](#)), ultrafast magnetism  
([Beaurepaire et al., 1996](#)), and ferromagnetic and semiconductive systems exhibiting anisotropic  
damping ([Chen et al., 2018](#)). A further set of applications stems from the extension of SpiDy.jl

42 to handle the non-Markovian stochastic dynamics of harmonic oscillators. This model will be  
 43 of interest in the field of quantum thermodynamics where harmonic oscillators play a key role  
 44 in modeling open quantum systems. The package is written in pure Julia to take advantage of  
 45 the language performance.

46 The software package has seen a wide range of applications to date. Firstly, the convenience  
 47 of three independent environments in SpiDy.jl finds application in the microscopic modeling of  
 48 spins affected by noise due to vibrations of the material lattice (Anders et al., 2022). SpiDy.jl  
 49 also found application in the demonstration of the quantum-to-classical correspondence at all  
 50 coupling strengths between a spin and an external environment (Cerisola et al., 2022). Here,  
 51 the temperature dependence of the spin steady-state magnetization obtained with SpiDy.jl is  
 52 successfully compared with the classical mean force state of the system. In Ref. (Hartmann  
 53 et al., 2023), the authors take advantage of the customizable coupling tensor in SpiDy.jl to  
 54 explore the anisotropic effects of the environment on the system. In Ref. (Berritta et al.,  
 55 2023), SpiDy.jl is used as a sub-routine to build quantum-improved atomistic spin dynamics  
 56 simulations. In the paper, the authors take advantage of the customizable power spectrum to  
 57 implement ad-hoc simulations matching known experimental results. Lastly, with an eye to the  
 58 harmonic oscillator side, SpiDy.jl is used to match the quantum harmonic oscillator dynamics  
 59 with its stochastic counterpart (Glatthard et al., 2023). Here, the authors exploit the recent  
 60 implementation of harmonic oscillator dynamics.

## 61 Overview

62 To model a system of interacting classical spin vectors, SpiDy.jl solves the generalized stochastic  
 63 LLG equation (Anders et al., 2022)

$$\frac{d\mathbf{S}_n(t)}{dt} = \frac{1}{2}\mathbf{S}_n(t) \times \left[ \sum_{m \neq n} J_{n,m} \mathbf{S}_m(t) + \mathbf{B} + \mathbf{b}_n(t) + \int_{t_0}^t dt' K_n(t-t') \mathbf{S}_n(t') \right], \quad (1)$$

64 where  $\mathbf{S}_n(t)$  represents the  $n$ -th spin vector, the interaction matrix  $J_{n,m}$  sets the interaction  
 65 strength between the  $n$ -th and  $m$ -th spins,  $\mathbf{B}$  is the external field, which determines the natural  
 66 precession direction and frequency of the spins in the absence of interaction, and  $b_n(t)$  is  
 67 the time-dependent stochastic field induced by the environment. Finally, the last integral  
 68 term in Eq.(1) gives the spin dissipation due to the environment, including non-Markovian  
 69 effects accounted for by the memory kernel matrix  $K_n(t)$ . Here, we allow each spin to interact  
 70 with three independent sources of noise so that in general  $K_n(t) = C_n k_n(t)$ , where  $k_n(t)$   
 71 is a time-dependent function and  $C_n$  is a  $3 \times 3$  matrix that determines how each of the  
 72  $n$ -th spin components couples to each of the three noise sources. To efficiently simulate  
 73 the non-Markovian effects, we follow the methods explained in (Anders et al., 2022) and  
 74 restrict ourselves to the case where the memory kernel  $k(t)$  comes from a Lorentzian spectral  
 75 density of the bath  $\mathcal{J}(\omega) = \alpha\Gamma/((\omega_0^2 - \omega^2)^2 + \Gamma^2\omega^2)$  with peak frequency  $\omega_0$ , peak width  
 76  $\Gamma$  and amplitude  $\alpha$ , so that  $k(t) = \Theta(t) \alpha e^{-\Gamma t/2} \sin(\omega_1 t)/\omega_1$ , where  $\omega_1^2 = \omega_0^2 - \Gamma^2/4$ . In  
 77 the code, the stochastic noise  $b_n(t)$  is generated so that it satisfies the fluctuation-dissipation  
 78 relation (FDR) (see (Anders et al., 2022)). That is, the power spectral density of the  
 79 stochastic noise satisfies  $P(\omega, T) = \mathcal{J}(\omega)\eta(T)$  where  $\mathcal{J}(\omega)$  is the Lorentzian spectral density  
 80 and  $\eta(T)$  defines the temperature dependence on the bath. Here, the user can choose, among  
 81 others, a classical or quantum-like temperature dependence, namely  $\eta_{cl}(T) = k_B T/2\hbar\omega$  and  
 82  $\eta_{qu}(T) = \coth(2\hbar\omega/k_B T)$  respectively.

83 In addition, SpiDy.jl also allows one to study the stochastic dynamics of coupled harmonic  
 84 oscillator networks. In the same way, as for the spin case, the harmonic oscillators can be  
 85 coupled together with a user-defined system-system interaction. The equations of motion  
 86 solved in this case are

$$\frac{d^2\mathbf{X}_n(t)}{dt^2} = \sum_{m \neq n} J_{n,m} \mathbf{X}_m(t) - \Omega^2 \mathbf{X}_n(t) + \mathbf{b}_n(t) + \int_{t_0}^t dt' K_n(t-t') \mathbf{X}_n(t'), \quad (2)$$

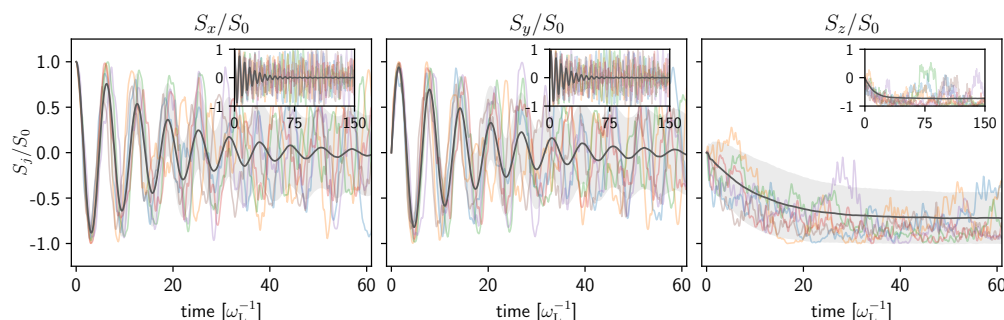
87 where  $\mathbf{X}_n(t)$  represents the position vector of the  $n$ -th harmonic oscillator, the interaction  
 88 matrix  $J_{n,m}$  sets the interaction strength between the  $n$ -th and  $m$ -th harmonic oscillators,  
 89 and  $\Omega$  is the bare frequency of the harmonic oscillators (we consider identical oscillators). All  
 90 other terms have the same role as in the spin case (see Eq.(1)).

91 In conclusion, SpiDy.jl implements the stochastic dynamics of coupled integro-differential  
 92 equations to model systems of interacting spins or harmonic oscillator networks subject to  
 93 environment noise. Among others, some of the key features of the package include:

- 94 ■ Coloured stochastic noise that satisfies the FDR and accounts for both classical and  
 95 quantum bath statistics.
- 96 ■ Simulation of non-Markovian system dynamics due to the memory kernel  $K_n(t)$ .
- 97 ■ Custom system-environment coupling tensors, allowing for isotropic or anisotropic cou-  
 98 plings. Both amplitudes and geometry of the coupling can be specified.
- 99 ■ Choice between local environments, i.e. distinct baths acting on the single sub-system,  
 100 or a single common environment.

101 In the next section, we show a minimal working example to run the spin dynamics where we  
 102 list all the required input parameters.

### 103 Example



**Figure 1: Single-spin dynamics.** Dynamics of the  $x$ ,  $y$ , and  $z$  spin components. The components are normalized against the total spin length  $S_0$  and time axes are expressed in units of the Larmor frequency  $\omega_L$  ( $\omega_L = |\mathbf{B}|$  in Eq.(1)). We show an example set of 5 stochastic trajectories of the spin dynamics (colored semi-transparent lines) together with their stochastic average (gray solid line). Note that, while we show only 5 trajectories for clarity, the average dynamics is obtained from 10000 trajectories. We also represent the range of one standard deviation from the average dynamics (gray-shaded area). In the inset, we show the convergence of the same dynamics towards the steady state at longer times. This example is obtained using the Lorentzian parameters “set 1” found in Ref. (Anders et al., 2022). The code used to generate the stochastic trajectories is shown in the text.

104 Now, we show an example code to generate a run of SpiDy.jl for a single spin interacting with  
 105 an environment. Given the stochastic nature of the problem solved, we will be dealing with  
 106 different dynamical trajectories. In the following code, we show the parameters needed to build  
 107 a subset of these trajectories, solutions to the stochastic differential equations of Eq.(1). Then,  
 108 we plot a single one of them as an example. The entire code is commented throughout for a  
 109 better understanding of the single elements of the run. We show the results of the dynamics  
 110 averaged over a larger set (10000) of trajectories in Figure 1. Note that both the average and  
 111 the standard deviation of the set of trajectories are not evaluated with the following code but  
 112 are nonetheless represented in the figure for clarity. Further examples are available in the code  
 113 repository.

```

### importing SpiDy ###
using SpiDy

### setting the parameters ###
ωL = 1 # Larmor frequency (reference time scale)
Δt = 0.1 / ωL # time step for the dynamics evaluation
tend = 150 / ωL # final time of the dynamics
N = round{Int, tend/Δt} # number of total steps
tspan = (0, N*Δt) # tuple of initial and final time
saveat = (0:1:N)*Δt # vector of times at which the solution is saved
α = 10 * ωL # Lorentzian coupling amplitude
ω0 = 7 * ωL # Lorentzian resonant frequency
Γ = 5 * ωL # Lorentzian width
Jsd = LorentzianSD(α, ω0, Γ) # Lorentzian spectral density
Cw = IsoCoupling(1) # isotropic coupling tensor
# the resulting coupling tensor is equivalent to the following
# Cw = AnisoCoupling([1 0 0
#                    0 1 0
#                    0 0 1]);
T = 0.8 * ωL # temperature at which the dynamics takes place (where ħ=1, kB=1)
noise = ClassicalNoise(T) # noise profile for the stochastic field
s0 = [1.0; 0.0; 0.0] # initial conditions of the spin vector for the dynamics
ntraj = 10 # number of trajectories (stochastic realizations)

### running the dynamics ###
sols = zeros(ntraj, 3, length(saveat)) # solution matrix
for i in 1:ntraj # iterations through the number of trajectories
    # we use the Lorentzian spectral density Jsd to generate the stochastic
    # field. This ensures the field obeys the FDR as noted in the main text
    local bfields = [bfield(N, Δt, Jsd, noise),
                    bfield(N, Δt, Jsd, noise), # vector of independent
                    bfield(N, Δt, Jsd, noise)] # stochastic fields
    # diffeqsolver (below) solves the system for the single trajectory
    local sol = diffeqsolver(s0, tspan, Jsd, bfields, Cw; saveat=saveat)
    sols[i, :, :] = sol[:, :] # store the trajectory into the matrix of solutions
end

### example plot ###
# use Plots.jl pkg to plot a single trajectory of the dynamics over time
using Plots
plot(xlabel="time", ylabel="spin components")
# sols[i,j,k] with i: trajectory index, j: spin component, k: solution at
# the k-th time point
plot!(saveat, sols[1,1,:], label="x-component")
plot!(saveat, sols[1,2,:], label="y-component")
plot!(saveat, sols[1,3,:], label="z-component")
savefig("example_run.pdf")

```

## 114 Acknowledgements

115 SS and FC thank Marco Berritta and Charlie Hogg for insightful suggestions for the implemen-  
 116 tation of the spin-spin coupling and the harmonic oscillator dynamics. SARH and JA thank the  
 117 Royal Society for their support. SS is supported by a DTP grant from EPSRC (EP/R513210/1).  
 118 SARH acknowledges the Royal Society and TATA for financial support through the Grant

119 URFR 211033. JA and FC acknowledge funding from EPSRC (EP/R045577/1). FC gratefully  
120 acknowledges funding from the Foundational Questions Institute Fund (FQXi-IAF19-01).

## 121 References

- 122 Anders, J., Sait, C. R. J., & Horsley, S. A. R. (2022). Quantum brownian motion for magnets.  
123 *New Journal of Physics*, 24(3), 033020. <https://doi.org/10.1088/1367-2630/ac4ef2>
- 124 Barker, J., & Bauer, G. E. W. (2019). Semiquantum thermodynamics of complex ferrimagnets.  
125 *Physical Review B*, 100(14). <https://doi.org/10.1103/physrevb.100.140401>
- 126 Beaurepaire, E., Merle, J.-C., Daunois, A., & Bigot, J.-Y. (1996). Ultrafast spin dynamics  
127 in ferromagnetic nickel. *Physical Review Letters*, 76(22), 4250–4253. <https://doi.org/10.1103/physrevlett.76.4250>
- 128
- 129 Berritta, M., Scali, S., Cerisola, F., & Anders, J. (2023). Accounting for quantum effects in  
130 atomistic spin dynamics. *arXiv Preprint*. <https://doi.org/10.48550/arXiv.2305.17082>
- 131 Cerisola, F., Berritta, M., Scali, S., Horsley, S. A. R., Cresser, J. D., & Anders, J. (2022).  
132 Quantum-classical correspondence in spin-boson equilibrium states at arbitrary coupling.  
133 *arXiv Preprint*. <https://doi.org/10.48550/arXiv.2204.10874>
- 134 Chen, L., Mankovsky, S., Wimmer, S., Schoen, M. A. W., Körner, H. S., Kronseder, M.,  
135 Schuh, D., Bougeard, D., Ebert, H., Weiss, D., & Back, C. H. (2018). Emergence of  
136 anisotropic gilbert damping in ultrathin fe layers on GaAs(001). *Nature Physics*, 14(5),  
137 490–494. <https://doi.org/10.1038/s41567-018-0053-8>
- 138 Ciornei, M.-C., Rubí, J. M., & Wegrowe, J.-E. (2011). Magnetization dynamics in the  
139 inertial regime: Nutation predicted at short time scales. *Physical Review B*, 83(2).  
140 <https://doi.org/10.1103/physrevb.83.020410>
- 141 Evans, R. F. L., Fan, W. J., Churemart, P., Ostler, T. A., Ellis, M. O. A., & Chantrell, R. W.  
142 (2014). Atomistic spin model simulations of magnetic nanomaterials. *Journal of Physics:  
143 Condensed Matter*, 26(10), 103202. <https://doi.org/10.1088/0953-8984/26/10/103202>
- 144 Glatthard, J., Hogg, C., Cerisola, F., & Anders, J. (2023). *In preparation*.
- 145 Halilov, S. V., Eschrig, H., Perlov, A. Y., & Oppeneer, P. M. (1998). Adiabatic spin dynamics  
146 from spin-density-functional theory: Application to fe, co, and ni. *Physical Review B*,  
147 58(1), 293–302. <https://doi.org/10.1103/physrevb.58.293>
- 148 Hartmann, F., Scali, S., & Anders, J. (2023). Anisotropic signatures in the spin-boson model.  
149 *arXiv Preprint*. <https://doi.org/10.48550/arXiv.2305.16964>
- 150 Neeraj, K., Awari, N., Kovalev, S., Polley, D., Hagström, N. Z., Arekapudi, S. S. P. K.,  
151 Semisalova, A., Lenz, K., Green, B., Deinert, J.-C., Ilyakov, I., Chen, M., Bawatna, M.,  
152 Scalera, V., d'Aquino, M., Serpico, C., Hellwig, O., Wegrowe, J.-E., Gensch, M., & Bonetti,  
153 S. (2020). Inertial spin dynamics in ferromagnets. *Nature Physics*, 17(2), 245–250.  
154 <https://doi.org/10.1038/s41567-020-01040-y>