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Optimal control of a quantum sensor: A fast algorithm based on an analytic solution

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Quantum sensors can show unprecedented sensitivities, provided they are controlled in a very specific, optimal way. Here, we consider a spin sensor of time-varying fields in the presence of dephasing noise, and we show that the problem of finding the optimal pulsed control field can be mapped to the determination of the ground state of a spin chain. We find an approximate but analytic solution of this problem, which provides a *lower bound* for the sensor sensitivity, and a pulsed control very close to optimal, which we further use as initial guess for realizing a fast simulated annealing algorithm. We experimentally demonstrate the sensitivity improvement for a spin-qubit magnetometer based on a nitrogen-vacancy center in diamond.

I. INTRODUCTION

Quantum systems are notoriously sensitive to external influences. This sensitivity is a core element in the development of quantum technologies, as is the case of quantum sensing, which takes advantage of quantum coherence to detect weak or nanoscale signals. Quantum sensing devices can in principle attain precision, accuracy, and repeatability reaching fundamental limits [1, 2]. However, the extreme sensitivity to external perturbations also causes the quantum sensor to couple with detrimental noise sources that induce decoherence, therefore limiting the interaction time with the target signal.

Here, we introduce a method to find optimal control protocols [3] for ac quantum sensing in the presence of dephasing noise. Such optimization problem is in general a complex classical problem. Our method, that draws an analogy between pulsed dynamical decoupling (DD) protocols [4–8] and spin glass systems [9], maximizes the phase acquired by the quantum sensor due to the target ac field while minimizing the noise detrimental effect. The optimal control fields yield an improved sensitivity, as we experimentally demonstrate using a spin-qubit magnetometer based on a Nitrogen-Vacancy (NV) center in diamond [10–14].

More in detail, we find that the problem of optimizing the control protocol for our quantum sensor

is homologous to that of finding the ground state of a classical Ising spin Hamiltonian. The control π -pulse times correspond to the locations in the chain of the domain walls. The couplings between the model spins are of both signs, and this is customary in optimization problems. The antiferromagnetic couplings capture the frustration between the different terms in the Hamiltonian, which then *prima facie* is that of a spin-glass model—which does not mean that there is a *spin-glass phase* at low temperature (see later).

The study of optimization problems in statistical physics is a large field of research in disordered systems, with far-reaching connections from the physics of spin glasses [15, 16] and other frustrated, classical and quantum models [17–23]. Optimization problems in quantum control can show some degree of frustration, with terms that compete in a similar way in which ferromagnetic and anti-ferromagnetic bonds compete in spin glasses [24]. We find, however, that in the specific case of the optimal control of a qubit sensor, by trading the Ising \mathbb{Z}_2 spins for the continuous spins of a spherical model (SM) [25] one gets rid of frustration altogether, and the model shows little signs of competing equilibria at low temperature, typical of replica-symmetry-broken phases [9, 26]. Since the ground state of the spherical model can be found analytically if the spectra of the signal and of the noise are known, we obtain from this both a *lower bound* for the sensitivity [27], and a quasi-optimal controlled pulsed field. The quasi-optimal sequence can be instead fed to a simulated annealing (SA) algorithm [28–31], in order to find the optimal

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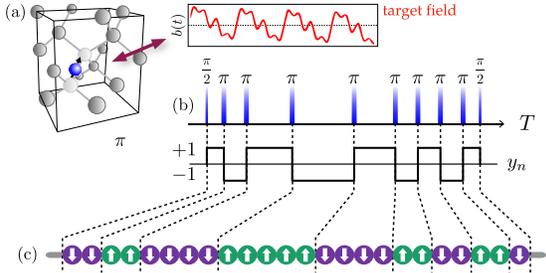


FIG. 1. (a) A single spin sensor is used to detect an AC target magnetic field $b(t)$. (b) An optimal control field applied to the spin sensor increases its coherence, hence improving its sensitivity. (c) The difficult problem of finding an optimal control sequence can be mapped into a problem of finding the ground state of a virtual spin chain.

one. Such annealed sequences show, in agreement with the experiments, very good sensitivities (only about 20% higher than the bound). To show the unparalleled performance of the algorithm, we also run it on a Raspberry Pi microcomputer, where it takes milliseconds to find the optimal solutions.

II. RESULTS

A. Optimized dynamical decoupling for sensing

We consider a single spin-qubit sensor of time-varying magnetic fields, in the presence of dephasing noise. This quantum sensing task can be described as a compromise between spin phase accumulation due to the external target field to be measured $b(t) \equiv bf(t)$, and refocusing of the non-markovian noise, obtained via dynamical decoupling (DD) protocols [4–8]. As in the Hahn’s echo [32–34], a DD sequence is implemented by applying sets of n π -pulses that act as time reversal for the phase acquired by the qubit during its free evolution, and can be described by a modulation function $y: [0, T] \ni t \mapsto \{-1, 1\}$ (see Fig. 1b). The DD sequence is embedded within a Ramsey interferometer, hence the qubit coherence is mapped onto the probability of the qubit to populate the excited state $|1\rangle$:

$$P(T, b) = \text{Tr}[\rho|1\rangle\langle 1|] = \frac{1}{2} \left(1 + e^{-\chi(T)} \cos \varphi(T, b) \right). \quad (1)$$

Here, φ is the phase acquired by the qubit during the sensing time T :

$$\varphi(T, b) = b\gamma \int_0^T dt f(t) y(t), \quad (2)$$

with γ the coupling to the field (e.g., the electronic gyromagnetic ratio of the spin sensor). The noise-

induced decoherence function

$$\chi(T) \equiv \frac{1}{\pi} \int \frac{d\omega}{\omega^2} S(\omega) |Y(T, \omega)|^2 \quad (3)$$

is the convolution between the noise spectral density (NSD) $S(\omega)$ and the filter function $Y(T, \omega) = i\omega \int_0^T dt e^{-i\omega t} y(t)$. Note that we neglect the effect of the target field on the noise source [35].

Dynamical decoupling is a very versatile control technique, with a virtually infinite space of degrees of freedom spanned by all the possible distributions of π pulses, even at finite sensing time T . One of the most common DD sequences is the Carr-Purcell (CP) sequence [33, 34], formed by a set of equidistant pulses. Non-equidistant sequences have been proposed and experimentally tested, e.g. in Refs. [7, 36, 37]. Each of these sequences has internal degrees of freedom, that can be tuned to increase the sensing capabilities for specific target fields. However, as the complexity of the target field increases, increases also the difficulty to find a pulse sequence that successfully filters out the noise components, while still maintaining the sensitivity to the target field.

A possible approach is to use an optimization algorithm to find a sequence that optimizes a desired figure of merit, for example the sensitivity, i.e. the smallest detectable signal. This concept was proposed and demonstrated experimentally for an NV center used as a quantum magnetometer [38]. The sensitivity, which can be expressed as (see Supplementary Information for a derivation)

$$\eta = \frac{e^{\chi(T)}}{|\varphi(T)/b|} \sqrt{T}, \quad (4)$$

encompasses both the effect of the detrimental noise, and the effect of the target ac field. Despite the achieved improvements [38], the computational complexity of the optimization problem limited its applicability.

In our approach, instead, the cost function η is re-defined to mimic the Hamiltonian of a classical Ising spin system. In this way, the continuous optimization problem for the minimization of the sensitivity of a NV-center magnetometer is re-interpreted as a discrete energy minimization problem. Specifically, we define the new cost function to be the (dimensionless) logarithmic sensitivity

$$\epsilon = \log \left(\eta \gamma \sqrt{T} \right) = \chi(T) - \log \left| \frac{\varphi(T)}{T\gamma b} \right|, \quad (5)$$

and we will show in Sec. II B that upon time discretization ϵ will become the Ising Hamiltonian, albeit with sign-alternating, long-range interactions and a peculiar logarithmic field-spin coupling. Before doing that, however, we show how the problem can be tackled with continuous time, and by means of a reasonable approximation.

B. A variational approach

Our task is to find the optimal function $y(t)$ which minimizes the sensitivity η —or the logarithmic sensitivity ϵ . First of all, we rewrite ϵ as

$$\epsilon[y] = \frac{1}{2} \int_{[0,T]^2} dt dt' y(t) J(t, t') y(t') - \log \left| \int_0^T dt h(t) y(t) \right|, \quad (6)$$

with

$$J(t, t') = \frac{2}{\pi} \int d\omega \cos(\omega(t' - t)) S(\omega). \quad (7)$$

Notice that J is a positive operator even though $J(t', t)$ can take up any values in \mathbb{R} , and we have the constraint $y(t)^2 = 1$ for all t . We can impose the constraint via a Lagrange multiplier function $\lambda(t)$:

$$F[y, \lambda] = \epsilon[y] + \frac{1}{2} \int_0^T dt \lambda(t) (y(t)^2 - 1). \quad (8)$$

We need to find the stationary point of $F[y]$ w.r.t. $y(t)$ and $\lambda(t)$. Formally, the saddle point equations are

$$\begin{aligned} \frac{\delta F}{\delta y(t)} &= \int_0^T dt' [J(t, t') + \lambda(t)\delta(t - t')] y(t') \\ &\quad - \frac{h(t)}{\int_0^T dt h(t) y(t)} \\ &= 0, \end{aligned} \quad (9)$$

$$\frac{\delta F}{\delta \lambda(t)} = y^2(t) - 1 = 0. \quad (10)$$

One can see that the extreme w.r.t. λ simply gives the constraint. The formal solution of the above equations is

$$y(t) = \frac{1}{D} \int_0^T dt' \left(\frac{1}{J + \lambda} \right)_{t, t'} h(t'), \quad (11)$$

where λ stands for the diagonal operator $\lambda(t)\delta(t - t')$, and

$$\begin{aligned} D &= \int_0^T dt h(t) y(t) \\ &= \frac{1}{D} \int_0^T dt dt' h(t) \left(\frac{1}{J + \lambda} \right)_{t, t'} h(t'), \\ \Rightarrow D &= \left(\int_0^T dt dt' h(t) \left(\frac{1}{J + \lambda} \right)_{t, t'} h(t') \right)^{1/2}. \end{aligned} \quad (12)$$

By plugging Eq. (11) in Eq. (8), one can express the cost function at the saddle as

$$\begin{aligned} F &= \frac{1}{2} \int_{[0,T]^2} dt dt' y(t) J(t, t') y(t') - \log \left| \int_0^T dt h(t) y(t) \right| + \frac{1}{2} \int_0^T dt \lambda(t) (y(t)^2 - 1) \\ &= \frac{1}{2D^2} \int_{[0,T]^2} dt dt' h(t) \left(\frac{1}{J + \lambda} \right)_{t, t'} h(t') - \log \left| \frac{1}{D} \int_0^T dt h(t) \left(\frac{1}{J + \lambda} \right)_{t, t'} h(t') \right| - \frac{1}{2} \int_0^T dt \lambda(t) \\ &= \frac{1}{2} - \log |D| - \frac{1}{2} \int_0^T dt \lambda(t). \end{aligned} \quad (13)$$

The last expression is a function of $\lambda(t)$ only and one can, in principle, find the saddle point of it and substitute it in (11) to obtain the optimum DD sequence.

Short of solving exactly the model in Eq. (8), we can get good results to guide the experiment by simplifying the space in which we are searching for the

minimum. We can do this in two ways: either we keep $y(t)$ defined on \mathbb{R} (i.e. we keep the time continuum) and we give more structure to $\lambda(t)$, or we discretize time and enforce the constraint $y(t)^2 = 1$ exactly (therefore getting rid of λ). These two approaches will be implemented in the following.

Spherical approximation. In order to make

progress, we substitute for the moment the constraint $y(t)^2 = 1$, for all t , with the constraint

$$\frac{1}{T} \int_0^T dt y^2(t) = 1. \quad (14)$$

This is equivalent to finding the stationary point of $F[y, \lambda]$, Eq. (8), in the subspace in which $\lambda(t) \equiv \lambda$. We call the resulting approximation *spherical model* (SM) since the constraint in Eq. (14), after discretization of time in N different, equally spaced values $t_i = i\Delta t$, puts the variable $y(t)$ on a sphere of radius $N = T/\Delta t$. The SM in the continuum limit, obtained by substituting $\lambda(t) \rightarrow \lambda$ in Eq. (13), provides one theoretical lower bound for the sensitivity: this simply follows from the fact that we are minimizing ϵ w.r.t. $y(t)$ over a *larger* space of functions.

Spherical models are often good mean field models of spin glasses and of their dynamics [25, 39], and this case will prove to be of similar nature despite the unusual logarithmic field coupling term. By setting $\lambda(t) \rightarrow \lambda$ we have the function of the single parameter

$$\epsilon_{\text{SM}}(\lambda) = -\frac{1}{2} \log \left| \int_0^T dt dt' h(t) \left(\frac{1}{J + \lambda} \right)_{t,t'} h(t') \right| + \frac{1}{2} - \frac{T}{2} \lambda, \quad (15)$$

where $J + \lambda$ is the operator with integral kernel $J(t', t) + \lambda \delta(t' - t)$. Maximizing w.r.t. $\lambda \in \mathbb{R}$, one finds a lower bound on the sensitivity: $\eta > \eta_{\text{SM}} = e^{\epsilon_{\text{SM}}} / \gamma \sqrt{T}$. In principle the bound is not sharp, however it provides a quick and accurate measure of the goodness of our results. Moreover, we have found by experience that it is *in practice* pretty close to being sharp and that it can hardly be improved analytically by adding more freedom to the function $\lambda(t)$ beyond the constant $\lambda(t) = \lambda$. For example the test function $\lambda(t) = \lambda_1 \chi_{[0, T/2]}(t) + \lambda_2 \chi_{[T/2, T]}(t)$ ($\chi_{[a,b]}$ is the characteristic function of the interval $[a, b]$), giving a two-parameters space (λ_1, λ_2) for minimization, gives at most a few percent increase on the bound on η . We therefore use it *as if it were sharp*.

One can, in fact, define for any DD sequence the dimensionless quantity $\eta_{\text{SM}}/\eta < 1$. We will see in the next section how different approximate solutions give different values of this quantity. Moreover, we will see how the solution of the SM, although not *per se* a DD sequence, can function as a starting point for finding an optimal DD sequence.

Time discretization and Simulated Annealing. Let us focus now on the second method: time discretization. We discretize the sensing time T into small time intervals Δt , to obtain a sequence of times $t_i = i\Delta t$ with $i \in 1, \dots, N = T/\Delta t$. The interval Δt is the smallest time we allow the π -pulses of the DD

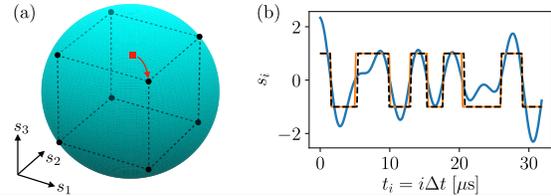


FIG. 2. (a) The solution of the spherical model (red square) is, in general, not a point on the hypercube (black dots), but it can be projected (arrow) onto the latter, giving a good value for the sensitivity. (b) Comparison between solutions for 200 spins ($T = 32 \mu\text{s}$, $\Delta t = 0.16 \mu\text{s}$): The continuous spins $s_i \in \mathbb{R}$ (blue line) can be converted into Ising spins $s_i = \pm 1$ (necessary for π -pulses) by using the sign function (orange line). The sensitivity can be improved further with a few iterations of SA to get a close-by sequence (dashed black). In this example, the trichromatic target signal and the noise are equal to the ones used in the experiment (see text).

sequence to be separated by. Apart from the physical limit given by the experimental apparatus, which gives a possible Δt to choose, one does not expect to need in the optimal solution π -pulses separated by much less than the minimum period of $b(t)$, if it exists (the spectrum of $b(t)$ can extend up to infinite frequency).

As introduced before, the modulation function at each of these times is $y(t_i) = \pm 1$, which dictates the sign of the phase acquired by the spin qubit during the time interval $[t_i - \Delta t, t_i]$. We can therefore write the modulation function as

$$y(t) = \sum_{i=1}^{N=T/\Delta t} s_i \chi_{[(i-1)\Delta t, i\Delta t]}(t), \quad (16)$$

where $s_i = \pm 1$, and as before $\chi_{[a,b]}$ is the characteristic function of an interval $[a, b]$. Writing the modulation function in this way allows us to recast Eqs. (2) and (3) respectively as

$$\varphi(T) = T\gamma b \sum_{i=1}^N h_i s_i, \quad (17)$$

$$\chi(T) = \frac{1}{2} \sum_{i,j=1}^N J_{ij} s_i s_j \quad (18)$$

where

$$h_i = \frac{1}{T} \int_{(i-1)\Delta t}^{i\Delta t} dt f(t) \quad (19)$$

represents the interaction with a normalized target ac field, and

$$J_{ij} \equiv \frac{4}{\pi} \int d\omega \frac{[1 - \cos(\omega\Delta t)]}{\omega^2} \cos(\omega(j-i)\Delta t) S(\omega) \quad (20)$$

represents the interaction with the detrimental noise. We can now express the new cost function as

$$\epsilon = \frac{1}{2} \sum_{i,j=1}^N J_{ij} s_i s_j - \log \left| \sum_{i=1}^N h_i s_i \right|, \quad (21)$$

that closely resembles the Hamiltonian of the Ising spin glass problem for a set of N spins s_i . The ground state for this Hamiltonian can be used to obtain a modulation function, therefore a DD sequence, that minimizes the sensitivity η .

At first sight, minimizing ϵ in Eq. (21) on the hypercube $\{s_i\} \in \{-1, 1\}^N$ can be in general a difficult problem, since the couplings J_{ij} can be of both signs. Therefore, one is tempted to use a simulated annealing (SA) minimization algorithm [28–30] to find the minimum of the energy ϵ . However, the performance of SA is strongly affected by the starting configuration in both in the final value and, at least as importantly, in the time to reach it. With this in mind we turn to the SM solved in the previous section but with our discretized time, in terms of which the spherical constraint reads $\sum_{i=1}^N y_i^2 = N$. In the discretized form, the solution of the SM is (see Eq. (11))

$$y_j = \frac{1}{D} \sum_{k=1}^N \frac{e^{i \frac{2\pi j k}{N}} \tilde{h}_k}{\sqrt{N} \tilde{J}_k + \lambda}. \quad (22)$$

Above, we introduced the Fourier transform of the signal term $\tilde{h}_k = \frac{1}{\sqrt{N}} \sum_j e^{-i \frac{2\pi k}{N} j} h_j$, and of the noise term $\tilde{J}_k = \frac{1}{\sqrt{N}} \sum_j e^{-i \frac{2\pi k}{N} j} J_{i,i-j}$: indeed, since $\lambda(t)$ is constant and J_{ij} depends only on the difference $i - j$, the matrix $J + \lambda$ is diagonal in Fourier space [40]. The value of λ is chosen to enforce the spherical constraint, and $D = \left(\sum_{k'} |\tilde{h}_{k'}|^2 / (\tilde{J}_{k'} + \lambda) \right)^{1/2}$. One can notice that in Fourier space the optimal solution is aligned with the field, and orthogonal to the noise.

An example solution is shown in Fig. 2. The values of y_i do not form a sequence of ± 1 , but the solution is reasonably close to the minimum of the original functional Eq. (21) over the hypercube $\{-1, 1\}^N$. We can now use the solution in Eq. (22) as a starting point to find the optimal sequence $s_i \in \{-1, 1\}$. To do so, we first define $s_i = \text{sign}(y_i) \in \{-1, 1\}$ and then run few steps of SA *moving only the domain walls*, i.e. flipping only spins which are on a sign change: $s_i = -s_{i+1}$. The π -pulse sequence is, as before, the sequence of times where the spins change sign (the position of the domain walls in the spin chain).

We test our procedure on an ensemble of test cases constructed as follows. The signal is a superposition $f(t) = \sum_{n=1}^{N_{\text{freq}}} A_n \cos(\omega_n t + \phi_n)$: we fix $N_{\text{freq}} = 7$ and extract uniformly random frequencies in the interval $[0, 1]$ MHz, uniformly random

phases ϕ_n , and uniformly random amplitudes A_n s.t. $\sum_{n=1}^{N_{\text{freq}}} A_n = 1$. The noise spectrum is instead a gaussian centered at 0.4316 MHz, and with standard deviation 0.016 MHz: thus, it is close to (but a little bit stronger w.r.t.) the experimentally relevant situation discussed in the next session.

First, we use a generalized Carr-Purcell (gCP) protocol in which the DD sequence is identified by the times at which the signal $f(t)$ changes sign, i.e. its zeros. This procedure is simple but not very effective: on average, it returns between 2/3 and 1/3 of the maximum inverse sensitivity, monotonically decreasing with the time of the sampling (see Fig. 3a). Second, we use the solution of the SM, viz. $s_i = \text{sign}(y_i)$, as DD sequence: this gives a better solution, but still not optimal. The best results are obtained by running a fixed number of steps of SA starting from the SM DD sequence $\text{sign}(y_i)$. As it is seen in Fig. 3a, this last strategy gives solution close to the upper bound given by the SM itself (before projecting on the hypercube).

Finally, we compare our results with an unbiased SA optimization, starting at infinite temperature from a uniformly random sequence of $s_i = \pm 1$. In this case, to reduce the number of π pulses it is necessary to introduce by hand a ferromagnetic coupling term in the Hamiltonian:

$$\epsilon \rightarrow \epsilon - K \sum_{i=1}^{N-1} s_i s_{i+1} \quad (23)$$

with $K > 0$ to be tuned. One can see in Fig. 3b that still the best sensitivity is obtained with the combination of the SM solution and SA optimization. Additionally, from Fig. 3b one can also understand that the optimal solution represents the best trade-off between number of π pulses (which the experimenter would like to maintain low) and sensitivity.

To conclude, we stress that our optimization procedure is very fast, if compared to standard, general-purpose routines. In particular, we were able to run our codes on a Raspberry Pi microcomputer, where the single instance takes ~ 0.5 s for the unbiased SA algorithm, and ~ 0.02 s for the solution of the SM and subsequent annealing (using $N = 500$ spins). This fact opens the door to the miniaturization of the control electronics, in view of possible technological applications of quantum sensing.

C. Experiment

While our method is general and applicable to any spin-qubit sensor, we exemplify it through experiments with a single NV center in bulk diamond with naturally abundant ^{13}C nuclear spins, at room temperature. The ground state electron spin of the NV

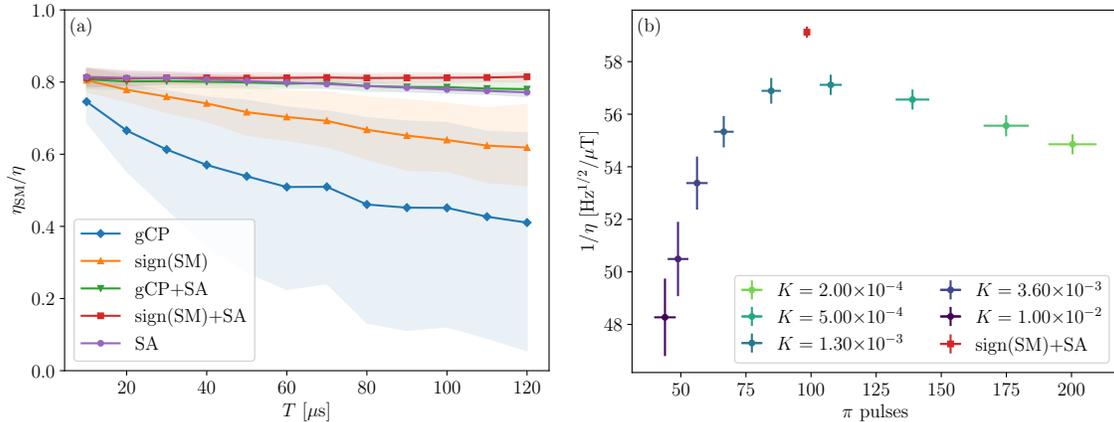


FIG. 3. (a) Comparison of performances for the DD sequences discussed in the main text: generalized Carr-Purcell (gCP), spherical model projected with the sign function onto the hypercube (sign(SM)), simulated annealing (SA), and SA optimization starting from gCP and sign(SM). One can see that the best results are obtained for the SA optimization guided by the SM solution. The data refer to the ensemble of random test signals described in the main text: the dots are the average values, and the shaded area represents the 20–80 percentile of the distribution of results. The discretization interval is $\Delta t = 0.1 \mu\text{s}$. (b) Single instance of a random signal, corresponding to $T = 100 \mu\text{s}$. It is shown the comparison of the unbiased SA, i.e. starting from infinite temperature (purple to green circles), and SA guided by the SM solution (red square). The unbiased SA needs a ferromagnetic term $\propto K$ —to be optimized over—in order to keep under control the number of π pulses (see Eq. (23)). From this plot, one learns that first, the optimal solution represents also the best trade-off between number of π pulses and sensitivity, and second, that the SA optimization guided from the SM performs better, and with less fluctuations. Here, each unbiased SA procedure uses 10^5 Monte Carlo steps and a power-law temperature ramp, while only 10^3 steps are needed for the SA from the SM solution. The dots represents the averages, and the error bars the standard deviations of the distributions.

center can be initialized and measured by exploiting spin-dependent fluorescence, and can be coherently manipulated by microwaves [14]. We consider the two ground-state spin levels, $m_S = 0$ and $m_S = +1$, to form the computational basis of the qubit sensor $\{|0\rangle, |1\rangle\}$ (see Supplementary Information). The main source of noise for the NV spin qubit derives from the collective effect of ^{13}C impurities randomly oriented in the diamond lattice.

In the presence of a relatively high bias field ($\gtrsim 150$ G), the collective effect of the nuclear spin bath on the NV spin is effectively described as a classical stochastic field, with gaussian noise spectral density (NSD) centered at the ^{13}C Larmor frequency ν_L [41, 42]. We preliminarily characterize the NSD of the NV spin sensor as in Ref. [42]. The direct coupling between the target field and the nuclear spins is negligible due to the small nuclear magnetic moment [35], and the indirect coupling via the NV electronic spin is also negligible due to the presence of the strong bias field [42]. Therefore, the NV spin dynamics is well described by Eq. (1). For the experiments we present throughout this article we used a bias magnetic field of $403.2(2)$ G, for which the NSD is $S(\omega) = S_0 + A \exp(-(\omega - \omega_L)^2/(2\sigma^2))$, with $S_0 = 0.00119(9)$ $\text{MHz}\cdot\omega_L/2\pi \equiv \nu_L = 0.4316(2)$ MHz ,

$A = 0.52(4)$ MHz , and $\sigma/2\pi = 0.0042(2)$.

As a test case for our optimal control method versus standard control, we consider a three-chromatic target signal, with $f(t) = \sum_{i=-1}^{+1} A_i \cos(2\pi\nu_i t)$, where $\nu_i = \{0.1150, 0.2125, 0.1450\}$ MHz are the frequency components, and $A_i = \{0.288, 0.335, 0.377\}$ are the relative amplitudes, respectively for $i = -1, 0, +1$.

In Fig. 4(a) we show the NV spin coherence $P(n\tau, b)$ under Carr-Purcell (CP)-type DD control, formed by n pulses with uniform interpulse spacing $\tau = T/n$, as a function of τ . The value of b at the position of the NV defect inside the diamond is obtained from minimizing the squared residuals between experiment (gray bullets) and simulation (gray line), for which b is the only free parameter (see Supplementary Information for more details).

The CP pulse sequence acts as a quasi-monochromatic filter centered at $1/\tau$, so that a single component of $b(t)$ can be sensed in each experimental realization. As a consequence, $P(n\tau, b)$ in Fig. 4(a) shows collapses occurring at $\tau \sim 1/2\nu_i$. Notice that the collapse corresponding to the frequency component ν_{+1} ($\tau \simeq 3.448 \mu\text{s}$) cannot be resolved from noise since the first harmonic of the filter function roughly coincides with the NSD peak

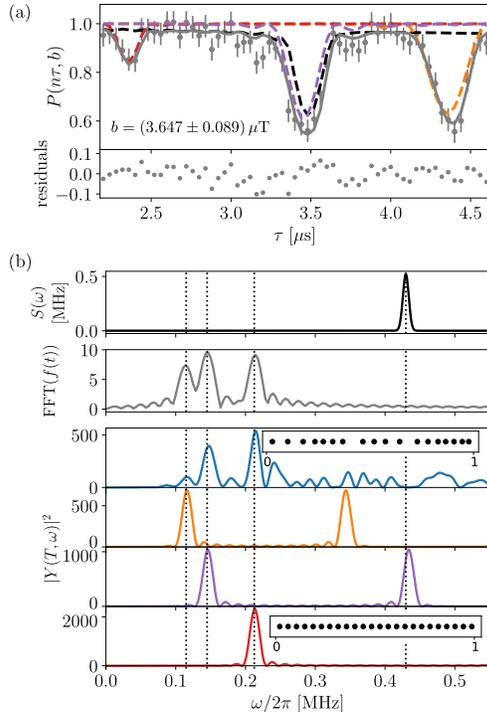


FIG. 4. (a) Dynamics of the NV spin qubit under a DD sequence with $n = 16$ equidistant pulses (CP) for a trichromatic signal (see text). The NV spin coherence is mapped onto the probability of the NV spin to be in the state $|1\rangle$, $P(n\tau, b)$. Gray bullets: experimental data. Black dashed line: simulated spin coherence in the presence of noise, without any external target signals. Orange, red, and purple dashed lines: simulated spin coherence in the presence of monochromatic target fields with ω_1 , ω_2 , and ω_3 , respectively, with no noise. Gray solid line: simulated data combining all of the above using Eq. (1). Residuals between gray experimental data and gray solid line are shown in the bottom plot. (b) NSD given by the nuclear spin environment of the NV sensor (black line); fast Fourier transform (FFT) of the target signal $f(t)$ (gray line). Vertical dotted lines: frequency components of the target signal, and center of the NSD. Orange, purple, and red lines: filter function for a CP sequence with $T = \frac{n}{2\nu_i}$, for $i = -1, 0$, and $+1$, respectively. Blue line: filter function of the optimized sequence. Inset: examples of time distribution of π pulses.

($\nu_{+1} \simeq \nu_L/3$) [Fig. 4(b)]. To detect the three components of the target signal and filter out the NSD, we need an optimized sequence. We thus apply the optimization algorithm detailed before to solve this experimental sensing problem.

In order to confirm the theoretical prediction on how the optimized DD sequence can outperform the standard control in terms of sensitivity, we performed measurements of the sensitivity itself. Specifically we used three different CP sequences, each

with time between pulses $\tau = \frac{1}{2\nu_i}$, for $i = -1, 0, +1$. Having a previous knowledge of the NSD allows us to predict the sensitivity of the the spin sensor using equations (3), (2), and (4), for any given DD sequence, and for any target AC signal $b(t)$. In Fig. 5a we show the estimated values for the inverse of the sensitivity as a function of the sensing time $T = n\tau$. Since $\tau = \frac{1}{2\nu_i}$ is fixed for each of the CP sequences, the variation of T corresponds to a variation of the number of pulses n . Notice how for $\tau = \frac{1}{2\nu_{+1}}$, the inverse of the sensitivity rapidly goes to zero. The estimated inverse sensitivity for the optimized sequences is also shown in Fig. 5a. The inverse sensitivity increases as a function of T , although we expect it to decrease at longer times due to decoherence.

In the experiment, we measure $P(T, b)$ as a function of the field amplitude b at a fixed sensing time. An example of this kind of measurements is shown in Fig. 5b. From the analysis of the oscillation of $P(n\tau, b)$, we can directly fit the values of χ and φ/b (see Eqs. (1) and (2)), and therefore we can obtain the values of η using Eq. (4). The sensitivity measured experimentally shows an excellent agreement with the expected simulated values (see Fig 5a). See Supplementary Information for two additional test cases: One for a monochromatic target signal such that the fifth harmonic of the NSD coincides with the frequency of the target signal; and one for a target signal with seven frequency components, all close to the NSD peak. These two cases confirm the results of the experiments shown in the main text.

III. DISCUSSION

We have shown that the problem of finding an optimal solution to quantum-control a single spin system for quantum sensing can be solved by first finding the ground state of a solvable spherical model of classical spins and then using this as a starting point for a simulated annealing algorithm. In this way this optimization algorithm is able to find a control sequence that shows a significant improvement to the sensitivity with respect to standard control sequences. In addition, from the spherical model we found a theoretical bound on the sensitivity. Although the spherical model can be mapped to a control sequence that gives relatively good results, using the simulated annealing algorithm is necessary to improve even further the sensitivity, approaching 80 – 85% of the bound. Our experimental results confirm the theoretical predictions, hence validating our algorithm as an optimization protocol applicable to single spin sensors.

The proposed algorithm can solve the problem of finding the optimal DD sequence of a given sig-

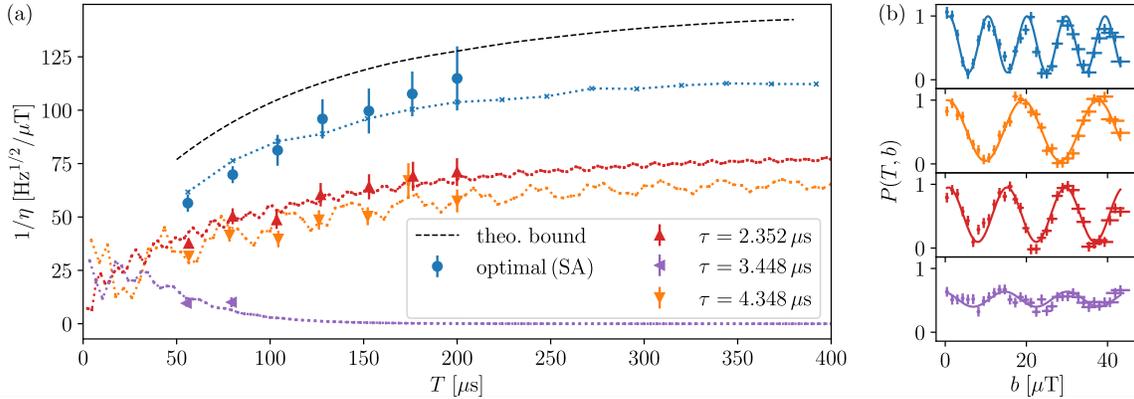


FIG. 5. (a) Experimental values of the inverse sensitivity for the optimized sequences (blue circles), and for the CP sequences (orange, red, and purple triangles). The predicted values of $1/\eta$ are represented by dotted lines. Black dashed line: theoretical upper bound of $1/\eta$, obtained from the solution of the spherical model in the continuum limit. (b) $P(T, b)$ as a function of the amplitude of b , at $T \simeq 152 \mu\text{s}$. Same color code as in (a). Lines are a cosine fit (see text).

nal $b(t)$ in a few *milliseconds* on a Raspberry Pi, which opens the door to miniaturization of the control electronics, using for example low-power processors. Fast optimization would also enable the implementation of adaptive protocols for sensing and

spectroscopy.

IV. DATA AVAILABILITY

The data that support the findings of this study are available from the corresponding author upon reasonable request.

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V. AUTHOR CONTRIBUTIONS

P.C., N.F., and S.H.-G. designed the experiment; S.H.-G. carried out the experiment and collected the data; G.F., F.B., and A.S. carried out the theoretical analysis and conceived the optimization protocol; S.H.-G. and F.B. analyzed the data. All authors have contributed equally to the writing of the manuscript.

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