Liquid-state nuclear magnetic resonance as a testbed for developing quantum control methods

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In building a quantum-information processor (QIP), the challenge is to coherently control a large quantum system well enough to perform an arbitrary quantum algorithm and to be able to correct errors induced by decoherence. Nuclear magnetic resonance (NMR) QIPs offer an excellent testbed on which to develop and benchmark tools and techniques to control quantum systems. Two main issues to consider when designing control methods are accuracy and efficiency, for which two complementary approaches have been developed so far to control qubit registers with liquid-state NMR methods. The first applies optimal control theory to numerically optimize the control fields to implement unitary operations on low-dimensional systems with high fidelity. The second technique is based on the efficient optimization of a sequence of imperfect control elements so that implementation of a full quantum algorithm is possible while minimizing error accumulation. This paper summarizes our work in implementing both of these methods. Furthermore, we show that taken together, they form a basis to design quantum control methods for a block-architecture QIP so that large system size is not a barrier to implementing optimal control techniques.

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I. INTRODUCTION

Quantum mechanics has been successfully used to describe experimental phenomena for close to a century. However, it was only much more recently that the idea of using quantum-mechanical evolution to process information was conceived [1]. This is revolutionizing the way we view the complexity of information processing. In particular, quantum-information processors (QIPs) may be able to perform certain computational tasks, such as factoring large integers or simulating quantum systems, exponentially faster than classical computers [2]. One of the difficulties in applying quantum mechanics to information processing is that quantum effects are extremely delicate and difficult to detect. The system must be well isolated from the environment, or decoherence will quickly destroy the quantum superpositions and entanglement that the QIPs exploit and reduce the system to a probabilistic classical state. It has only recently been realized that it is possible to efficiently encode quantum information to protect it from decoherence. As a result, the control and isolation of quantum-information systems need not be arbitrarily good. It suffices for the computationally relevant error probability per basic quantum gate to be \( \lesssim 10^{-4} \) [3] and with extreme overheads \( \lesssim 10^{-2} \) [4]. In practice, satisfying this error bound requires balancing the need for strong interactions to be able to apply quantum gates and good isolation to ensure that the quantum systems evolve coherently. Moreover, in the context of building a large-scale QIP, the classical resources required to implement the desired control must scale efficiently with the size of the QIP. This is an extraordinary challenge, but the prospect of being able to exploit the power of quantum computing has provided the impetus to explore many technologies in the pursuit of a scalable quantum computer [5].

One such technology is based on nuclear magnetic resonance phenomena. Nuclear magnetic resonance (NMR) phenomena are well described by quantum mechanics [51], and the control technology developed over its more than 50 years of existence has allowed for sophisticated pulse sequences that exploit the full range of the system dynamics. By leveraging NMR engineering experience and its well-developed control technology, NMR has been used to realize many of the first demonstrations of quantum algorithms.

As NMR quantum-information experiments increased in complexity, so did the quantum control methods used. In the first implementations of quantum algorithms on small molecules, such as the Deutsch-Jozsa [6] and the Grover’s [7] algorithms, preparation of the three-qubit GHZ state [8] and the implementation of the three-qubit quantum error correction code [9], the control sequences were written entirely by hand based on well-known techniques from NMR and heuristic optimization. In later more sophisticated experiments on larger molecules, such as a seven-qubit quantum superposition (cat state) experiment and the implementation of the five-qubit code [10,11], the control sequences were still based on standard NMR radio frequency (rf) shaped pulses, but, due to the large number of degrees of freedom, a computer compiler was built to systematically, numerically optimize the pulse sequences. Using similar techniques, VanderSypen et al. demonstrated the simplest instance of Shor’s factoring algorithm on a seven-qubit register [12]. At the same time, Cory’s group at MIT developed a different control approach based on the design of numerically optimized high-fidelity pulses that refocused all unwanted interactions [13]. More recently, both approaches have been used to control a 13-qubit system [14].

The availability of long-lasting coherence and high-fidelity control makes NMR an ideal testbed for quantum control techniques. In this paper we present two approaches to NMR quantum control that we have used. The first approach applies optimal control theory to achieve high fidelity and robust control over small spin systems. The second is a scalable technique, where a well-chosen approximation al-
allows for imperfect control elements to be placed in sequence while controlling the error buildup. We show how the two methods can be combined to apply optimal control methods to larger systems in a scalable manner. The sequence of basic operations needed to implement an algorithm can be described in the circuit model of quantum computing. In this model the sequence of operations is given by a quantum circuit, a time-ordered sequence of quantum gates to be applied to the quantum systems. A typical quantum circuit has three main components: initialization, implementation of the unitary evolution that realizes the desired algorithm, and a final measurement to obtain the algorithm’s output. In this paper we focus on the unitary evolution. We show how one can modulate the available control parameters to implement a universal set of quantum gates and how one can combine the gates to realize a complete quantum circuit.

The paper is divided into the following sections: In Sec. II the relevant aspects of the system and principles of control in an NMR QIP are explained; in Sec. III we describe both control methods: An optimal control method based on an exact description of the system, and an efficient approximation scheme based on subsystems; in Sec. IV, we explain why they form a basis for a flexible and general framework to design practical quantum control methods. More details about the technical implementations of such methods for liquid state NMR can be found in the appendices.

II. NMR

The basic principles of NMR are available in literature [15,16], and their use in quantum computing has been reviewed in [17–20]. Here we discuss only the relevant basics. In liquid-state NMR the qubits are the spin-$\frac{1}{2}$ nuclei of a molecule. The number of magnetically unique spin-$\frac{1}{2}$ in a molecule gives the number of qubits. A sample with approximately $10^{20}$ identical molecules dissolved in a liquid solvent averages to zero the pseudodipolar or anisotropic portion of the $J$ coupling and the dipolar coupling, both within the same molecule and (to a good approximation) between molecules [16]. This leaves only the weaker and isotropic $J$ coupling. The coupling Hamiltonian between spins $i$ and $j$ is of an exchange form,

$$\mathcal{H}_{J \text{ coupling}} = \frac{\pi}{2} J_{ij} (X_i X_j + Y_i Y_j + Z_i Z_j).$$  \hspace{1cm} (2)

Since this coupling is mediated by the electrons binding the two nuclei, there is no $J$ coupling between different molecules. If the difference in the resonant frequencies of the two nuclei involved is much greater than the coupling strength, then a secular approximation is valid. The off-diagonal terms can be ignored and the coupling is reduced to the weak-coupling Ising form where the $X_i X_j$ and $Y_i Y_j$ terms are dropped,

$$\mathcal{H}_{J \text{ coupling weak}} = \frac{\pi}{2} J_{ij} Z_i Z_j.$$ \hspace{1cm} (3)

This approximation is always valid for heteronuclear systems and for the majority of the homonuclear systems considered for a QIP. With this approximation, all of the terms in the internal Hamiltonian commute, and tracking the system during periods of free evolution becomes particularly straightforward. This can be used to simplify the control schemes.

The internal Hamiltonian enables one-qubit rotations about the $Z$ axis and two-qubit controlled-$Z$ gates. However, for universal control, one-qubit rotations about another axis are needed. These are implemented by control-field Hamiltonians. By applying an rf field, we can induce transitions between energy levels whose energy difference is resonant with the field. In the rotating frame the control-field Hamiltonian can be described as

$$\mathcal{H}_{\text{control}} = \omega_{\text{nut}}(t)(\cos[\phi(t)]X + \sin[\phi(t)]Y).$$ \hspace{1cm} (4)

Up to the hardware limitations of the spectrometer, it is possible to arbitrarily control the amplitude ($\omega_{\text{nut}}$; the nutation or Rabi frequency) and phase ($\phi$). With a standard liquid-state probe, nutation frequencies of up to 30 kHz are feasible. If the rf is on resonance with the spins’ Larmor frequency, then in the rotating frame the contribution to the Hamiltonian of...
the static magnetic field along Z vanishes and the rf field looks like a constant field about an axis \( \phi \) from the x axis. Hence, the spins precess about this axis at a frequency \( \omega_{\text{rf}} \), and we can induce rotations about any axis in the X-Y plane in the rotating frame. In the reference frame rotating at the rf, spins whose transition frequencies are off-resonant have an additional Z component in their Hamiltonian, and the effective rotation axis will be the vector sum of the rf field and the off-resonant Z field. If the spins are far off-resonance then this rotation axis will be close to the z axis and the spins will not be rotated into the plane. However, the spins are not completely unaffected and pick up an additional phase or z rotation known as the transient Bloch-Seigert shift \([21]\), which must be accounted for (see Sec. III B). The combination of the internal Hamiltonian and the control fields gives universal control, i.e., it is possible to implement any unitary transformation on the system. The challenge is to find the control fields as a function of time that drive the system through the desired evolution.

Some subtle effects arise from the fact that NMR QIPs utilize an ensemble of quantum systems, rather than a single one, as is commonly assumed in quantum computing:

State preparation. In the liquid state the sample temperature cannot vary much from room temperature. Even with the very strong fields available, the Zeeman energy splitting is much smaller than the available thermal energy and so the thermal state is highly mixed, with only a very small \( (\epsilon \sim 10^{-5}) \) bias toward the ground state. The thermal state for a single spin is of the form

\[
\rho_{\text{thermal}} = \frac{1}{2} + \epsilon \frac{1}{2} + \epsilon |0\rangle \langle 0|. \tag{5}
\]

In NMR quantum-information processing experiments, the evolution of the state is to a good approximation unital, meaning that the identity component of the state is preserved (the experiment time is much shorter than \( T_1 \)). In addition, the identity component is not directly observable because it has no effect on sample magnetization. For our purposes, it is sufficient to observe the dynamics of the second component of the state, called the deviation density matrix, which is pure. This component is called the pseudopure state. The idea of and method for preparing pseudopure states can be generalized to multiple qubits,

\[
\rho_{\text{pp}} = \frac{1 - \epsilon}{N} I + \epsilon |\psi\rangle \langle \psi|, \tag{6}
\]

where \(|\psi\rangle\) describes a pure quantum state on \( n \) qubits and \( N = 2^n \) is the dimension of the Hilbert space. Unfortunately, the standard preparation methods \([22-24]\) are inefficient and \( \epsilon \) decays exponentially with the number of qubits in a pseudopure state. Thus the signal quickly disappears in the noise, and this limits the use of traditional NMR methods to manipulate quantum information to less than approximately 15 qubits. Furthermore, because the initial state is so highly mixed, the states are provably separable, and a classical model, although somewhat contrived, can describe the results of NMR experiments \([25]\). Nevertheless, one can still use pseudopure states to explore quantum control issues. It should be emphatically stated that the efficiency of pseudopure states is not a fundamental limitation to the scalability of an NMR QIP. The spin temperature or polarization is not limited to the sample temperature and several approaches such as dynamic nuclear polarization \([26]\) or parahydrogen \([27]\) have been successful in creating very high spin polarizations in the liquid state. Furthermore, scalable, efficient algorithms have been developed to create pure states with NMR \([28,29]\). These algorithms are not yet practical with current technology in liquid-state NMR. However, they are practical in other magnetic resonance systems where high polarization is achievable, such as solid-state NMR \([30]\) and electron-spin resonance (ESR) \([31]\). We expect the control techniques developed here to readily transfer to these systems.

Measurement. In standard implementations of a QIP, the ability to make single-qubit projective measurements is assumed. However, in the ensemble framework of NMR we can measure only the expectation value of certain observables averaged over the ensemble of identical systems. Fortunately, most quantum algorithms can be modified to fit this framework \([23]\). Furthermore, the ensemble measurement is also sufficient to measure the fidelity of any unitary transformation applied to an initial pseudopure state. This makes it possible to benchmark control processes. There are some subtle issues regarding the coarseness of the measurement, the size of the system, and how much the system is disturbed by the measurement \([32]\); however, for typical samples these can be ignored.

Despite the problems due to the ensemble nature of the system, the resulting Hamiltonian and control principles are applicable to many different systems. Examples include other spin systems, such as solid-state NMR \([30]\), or electron-nuclear systems in organic crystals \([31]\), quantum dots \([33]\) or nitrogen-vacancy centers in diamond \([34]\), and even nonspin systems governed by effective spin Hamiltonians such as superconducting qubits \([35]\). The control principles developed should therefore find application to these other, potentially more practically scalable systems.

III. CONTROL METHODS

Standard approaches to control in NMR spin systems are well known \([18]\) and have permeated other quantum-information processing technologies. In the present work we describe two different approaches taken by our group to design and optimize pulse sequences, each addressing an important issue. First, given an accurate model of our system and apparatus, we can design high-fidelity control sequences by simulation and optimization. Second, we make some well-chosen approximations to our model to design accurate and robust sequences efficiently.

A. Optimal control theory

By applying standard optimal control theory methods to the problem of controlling quantum systems, high-fidelity control sequences can be found. The basic idea is to let some form of numerical optimization search for the pulse shape of
the control fields that exactly implements the desired unitary gate. The pulse will internally refocus all unwanted couplings and Bloch-Siegert shifts. Although we are left with little physical intuition about the path the system is driven through, very-high-fidelity control sequences are possible in a variety of situations where simple shaped pulses do not work. Two concerns with this powerful approach are its efficiency and practicality. In principle, since universal control is available, there exists a pulse shape that will implement any desired unitary. However, finding it requires a simulation of the full system which limits the optimal control approach to systems of less than approximately 8 qubits on a desktop computer and 10 qubits on larger clusters [36]. In addition, the pulse may require many hundreds if not thousands of time steps that need to be optimized, giving a large parameter space over which to optimize. A huge search space combined with expensive function evaluations leads to a difficult optimization problem.

Quantum optimal control is well studied in the context of driving chemical reactions with shaped laser pulses. However, this addresses a state transformation problem, whereas for quantum computing purposes, it is necessary to solve the much more difficult problem of realizing a unitary transformation that works for all input states. The total Hamiltonian is the sum of the rf control Hamiltonians and the internal Hamiltonian,

$$H_{tot}(t) = H_{int} + H_{rf}(t).$$  

(7)

As discussed in Sec. II, the internal Hamiltonian is of the form $H_0 = \sum_i \omega_i Z_i + \sum_i J_{ij} Z_i Z_j$. This gives us Z rotations and controlled-Z evolution, which is the basis of the two qubit gates. The internal Hamiltonian is time independent and its parameters can be obtained with high precision from standard spectroscopic techniques and fitting software. The free parameters that can be used to control the evolution of the system are in the rf Hamiltonian: $H_{rf}(t) = \sum_i u_i(t) \Sigma_i X_i + u_j(t) \Sigma_i Y_j$, where $k$ lists the spins sensitive to the rf pulse, which is the spins of the same nuclear species. The optimal control problem is to find the values of the control fields $u_k(t)$ that ensure that the total unitary evolution is as close as possible to the desired one. There will be many control-field sequences that give the correct evolution, but to minimize decoherence the time-optimal solution is preferred. For relatively simple systems with two or three qubits, and with a few potentially unrealistic assumptions on the experimentally available control, this can be solved analytically [37,38]. However, for bigger systems it is necessary to use a numerical optimization procedure to find control sequences. The algorithms start with a random guess for the control fields, which are discretized at a sufficiently high resolution in time. The system is then simulated and the unitary $U_{sim}$ is obtained. This is then compared with the goal unitary $U_{goal}$ through a well-chosen fitness function. Since global phases are irrelevant, a suitable choice that gives freedom in the global phase is the Hilbert-Schmidt (HS) fidelity, defined by

$$\Phi = \|\text{tr}(U_{goal}^\dagger U_{sim})\|^2 / N^2.$$  

(8)

This fidelity measure can be seen as an imperfect motion reversal and is linearly related to the average fidelity, defined as the squared state overlap between the ideal output state and the simulated output state averaged (Haar measure) over all input states [39]. Given this fitness function, any numerical optimization scheme can be used to search for the highest-fidelity pulse.

The first work on numerically optimizing rf control for an NMR QIP was performed by Cory’s group at MIT [13]. Their approach was to simplify both the search space and the function evaluations by using pulses consisting of blocks of constant amplitude, phase, and frequency. Approximately 10 to 30 of these blocks were combined into a composite pulse called a “strongly modulating pulse.” Thus, there were only 10 to 30 time steps to evaluate and compose in order to compute the overall unitary. The parameter space of the composite pulse was then searched with a simplex search algorithm. This method successfully found high-fidelity pulses that were experimentally implemented in a variety of quantum-information processing demonstrations [40]. However, the search method is not optimal and has difficulty finding high-fidelity pulses for more than four qubits. In addition, the pulses have sharp discontinuities between the blocks that lead to transient effects in pulse generation and in the NMR probe resonant circuit and suboptimal experimental implementation.

A substantial improvement was made by applying standard optimal control theory to the problem. This resulted in the gradient ascent pulse engineering (GRAPE) algorithm of Khaneja et al. [41,42]. If a pulse is digitized the control fields become a sequence $u_j(t)$ where $j = [1, \ldots , M]$ denotes the $j$th time step. The unitary for each time step is,

$$U_j = \exp \left[-i \Delta t (H_{int} + \sum_i u_i(j) H_i) \right],$$  

(9)

where $H_i = \Sigma_i X_i$, for example, and $\Delta t$ is the length of the time step. To first order, the derivative of the unitary propagator with respect to the control fields can be evaluated without finite differencing or another matrix exponential as

$$\frac{\delta U_j}{\delta u_i(j)} = -i \Delta t H_i U_j,$$  

(10)

where we require $|\Delta t H_i| < 1$ for the approximate derivative to be accurate. The total unitary for the pulse can be calculated as the product of each time-step unitary,

$$U_{tot} = U_M U_{M-1} \cdots U_2 U_1.$$  

(11)

The gradient of the fitness function, Eq. (8), can now be explicitly calculated as

$$\frac{\delta \Phi}{\delta u_i(j)} = \frac{1}{N^2} \text{tr} \left( (U_{j+1} \cdots U_M U_{goal})^\dagger \frac{\delta U_j}{\delta u_i(j)} U_{j-1} \cdots U_1 \right),$$  

(12)

Thus by storing the forward and backward propagation of the unitary and substituting Eq. (10) into Eq. (12), gradient information about the fitness function can be obtained without finite differencing or recalculation of the entire propagator. The gradient leads to a much more efficient search determin-
The numerically optimized control sequences drive the performance of the pulse restricted to each subsystem. If the control sequence implement the same desired unitary for a range of control powers. Robustness with respect to both the rf field strength and static field. Thus, the performance of the pulse is unaffected by inhomogeneities or miscalibration. This robustness also obviates the need to find new control pulses due to small changes in the internal Hamiltonian over time.

An important advantage of the GRAPE method is that the more efficient search method allows the pulse to be defined with many more points so that smooth “spectrometer friendly” pulses can be obtained, as explained in Appendix B. Our code is based on the algorithm in [41] with some modifications also discussed in the appendix. One important modification is the ability to define subsystems for a particular molecule and to define a fitness function as a sum of fitness functions [Eq. (8)] restricted to the subsystems. If the desired unitary is a tensor product across the subsystem boundaries then $U_{\text{goal}}$ can be defined for each subsystem and the performance of the pulse restricted to each subsystem calculated. This limits the size of spin systems which need to be simulated and dramatically speeds up finding pulses. The code finds high fidelity (above 99.75% HS fidelity including experimental inhomogeneities in rf and static fields) in a number of systems with up to seven qubits. The optimization method finds only local maximum in the search space so that better pulses may exist, but the local maxima found have proven to be sufficient for high-fidelity control.

The numerically optimized control sequences drive the system through a complicated and nonintuitive path. As a result it is possible that small errors in the model of the system and apparatus could lead to a much lower fidelity pulse. Fortunately, the optimization method can be modified so that the pulses are robust in the presence of static inhomogeneities in the system and the control Hamiltonian. For example, due to mismatch of magnetic susceptibilities of the sample and sample tube and imperfect shimming, the static magnetic field varies across the sample. We can demand that the control sequence implement the same desired unitary for a range of static magnetic fields, so that the pulse is robust given the corresponding variations in the Larmor frequencies. Another practical example is that the amplitude of the control fields may not be perfectly calibrated and again vary across the sample because of coil geometry, so we can also demand that our pulse implement the desired unitary for a range of control powers. Robustness with respect to both these effects for a particular seven-qubit pulse is shown in Fig. 1.

**B. Efficient presimulation and optimization**

In larger systems an alternative approach to control must be taken because the classical cost of simulating the full quantum dynamics grows exponentially with the system size. This prohibits the application of optimal control techniques to the full system. Some well-chosen simplifications and approximations must be made to the model. The control sequence is constructed from simple predefined but imperfect
building blocks (for example, standard frequency selective pulses). The blocks are then systematically put together to form a pulse sequence ensuring that the errors in the building blocks do not build up excessively as the sequence progresses. It is possible to efficiently design such sequences only if we judiciously simplify the model and take into account only the largest and first-order errors in the blocks.

Typical building blocks are one-qubit unitaries that involve selectively rotating one spin. If the spins have distinct resonant frequencies, this corresponds to frequency selective pulses. The problem of obtaining such pulses has of course had much attention in the long history of NMR. The most successful approach has been to use shaped pulses, and a huge variety of increasingly complicated shaped pulses have been developed with various bandwidths, excitation profiles (tipping angles as a function of chemical shift), and robustness to experimental imperfections [43]. In the linear regime the excitation profile in frequency space of a pulse is related to its Fourier transform [52]. As one would expect, the longer the pulse, the more selective it is in frequency space. Furthermore, one can tailor the excitation profile by shaping the pulse. For example, a Gaussian shaped pulse has a Gaussian shaped excitation profile. Thus, given the internal Hamiltonian of the molecule, it is straightforward to design a set of pulses for single spin rotations. An example of a pulse and its excitation profile is shown in Fig. 2.

For many NMR spectroscopy experiments, it is sufficient to consider only the excitation profile, but for quantum computing purposes we need to accurately keep track of all the effects of the pulse. In particular there are off-resonant and coupling effects. Although the resonant frequency of a particular spin may be far outside the excitation bandwidth of a selective pulse intended for another spin (so that there is no rotation about an axis in the plane), there still is a substantial phase evolution from the off-resonant or transient Bloch-Siegert effect [21]. To first order, the effect of an off-resonant pulse is to shift the resonant frequency of the spins, inducing an extra phase rotation. For example, a spin 3 kHz off-resonant from a 1 ms pulse that performs a 180° rotation on an on-resonance spin will experience an additional phase shift of $\sim 15^\circ$. This is non-negligible and must be accounted for. Second, the couplings still evolve during the pulse (which may have a duration comparable to $1/2|J|$), and the coupling Hamiltonians do not commute with the rf Hamiltonian. Particularly for long pulses on spins that have strong couplings, there are substantial deviations from the ideal action of the pulse. These can be partially countered by self-refocusing pulses [43], such as the Hermite pulses (see Table II).

Both the off-resonant and the coupling effects can be accounted for through a decomposition scheme where the imperfections are unrolled from the simulated pulse and represented as phase and coupling errors before and after the ideal pulse. A similar method with symmetric negative time evolution was presented in Ref. [19]. The method described here is more accurate and general. The simulated pulse is modeled by the following decomposition:

$$U_{\text{sim}} = e^{-i[H_{\text{rf}} + \sum_i \delta_{i}^{\text{post}} Z_i + \sum_{i<j} J_{ij} Z_i Z_j] \Delta t}} \prod_{i} e^{-i\delta_{i}^{\text{post}} Z_i} \prod_{i<j} e^{-i\delta_{ij}^{\text{post}} Z_i Z_j} U_{\text{ideal}} \prod_{i} e^{-i\delta_{i}^{\text{pre}} Z_i} \times\prod_{i<j} e^{-i\delta_{ij}^{\text{pre}} Z_i Z_j}.$$  

(13)

This decomposition resolves the simulated pulse into the ideal unitary operator sandwiched on either side by Z rotations (to account for the Bloch-Siegert shift and the chemical shift evolution during the pulse) and ZZ couplings to account for the couplings that occurred during the pulse (see Fig. 3). This model for the pre-error and post-error terms does not work for arbitrary pulses, but in the relevant case of spin selective pulses, it will capture most of the first-order dynamics. All couplings that do not involve the target qubit are trivial to extract and can be perfectly reversed. For couplings involving the target qubit, the control fields modulate the coupling necessitating a numerical optimization of the error terms. For 90° pulses ZZ pre-error and post-error terms can represent coupling evolution during the pulse. However, for 180° pulses, because the ideal pulse should refocus the couplings, ZZ error terms are not sufficient, and self-refocusing pulses are desirable. Since the pre-error and post-error terms
do not commute with the control Hamiltonian, the decomposition faithfully represents the true dynamics only when $|\Delta t/J| \ll 1$ where $\Delta t$ is the length of the pulse and $J$ is the strongest coupling.

The optimal parameters for the $Z$ and $ZZ$ “error” terms can be determined by use of an efficient procedure. A set of single-spin and pairs-of-spins simulations is sufficient to capture off-resonant effects and some first-order coupling contributions, since the Hamiltonian contains only one- and two-body terms, and all the pre-error and post-error terms commute with each other. The simulations are of a fixed one- or two-qubit size, hence there are only $n + \frac{n(n-1)}{2}$ simulations to perform. From these simulations, the optimal pre-error and post-error terms are determined by use of a numerical search. See Appendix C for a detailed description of how to obtain the error terms. The key points are that an efficient approximate description of the pulse is possible, and that the deviations from the ideal pulse in the decomposition can be corrected later as part of a larger pulse sequence by means of periods of free evolution and individual rotating frames adjustments (as shown in Fig. 3). An example decomposition of the pulse implementing a 90° rotation on $C_2$ (shown in Fig. 2) is given in Table I. The representation of the pulse as Eq. (13) has an average fidelity of 99.96% with respect to a simulation of the full system dynamics.

These pulse representations give us the building blocks for one-qubit gates. Two-qubit gates are achieved by periods of free evolution during which the coupling terms in the Hamiltonian evolve. If a coupling term is allowed to evolve for a time $1/2J$, then a controlled-$Z$ gate (up to one-qubit $Z$ rotations) is implemented between the coupled spins. All undesired couplings must be refocused with 180° pulses. Setting the pulses and delays together to form a pulse sequence that implements a desired algorithm has been automated through the use of a sequence compiler.

Several further simplifications are used. Individual rotating frames are defined for each spin. Just as the rotating frame for a single spin eliminates the Zeeman term in the Hamiltonian, individual rotating frames eliminate all chemical shift terms from the internal Hamiltonian. This could be implemented by a transmitter dedicated to each spin and rotating at that spin’s chemical shift, but this is experimentally prohibitive. The phase evolution caused by the chemical shift

![Diagram of pulse decomposition](image-url)
term in the Hamiltonian is simply a Z rotation, and this evolution can be tracked with respect to some fixed reference frame. Similarly the evolution of the transmitter at its frequency can be tracked, and when a pulse is required, the pulse phase is adjusted to obtain the correct phase in the spin’s rotating frame. Similarly the observation phase must also be calculated and adjusted. This tracking greatly simplifies the pulse sequences by eliminating the need to refocus the chemical shift evolution. It also allows Z rotations to be implemented instantaneously and with high precision through a frame change—that is, by simply modifying the phases of subsequent pulses and potentially the observation phase. Thus it is worth transforming the quantum circuits to use Z rotations preferentially. A drawback of individual reference frames is that it is not always possible for the pulse to have the correct phase in two different reference frames, so global pulses that nontrivially affect multiple spins are not always feasible, and hence they must be decomposed into a sequence of simpler pulses.

The coupling gates are implemented with delay periods to allow the desired couplings to evolve. Unfortunately, the undesired couplings also evolve. The unwanted couplings can be refocused with a series of 180° pulses by means of standard, efficient algorithms [44,45]. However, it is not necessary to refocus all unwanted couplings during every evolution period. It is sufficient to track the evolution, and refocus it only when needed [46]. Single qubit gates and the Z and ZZ error terms in their representation commute with the terms of the internal Hamiltonian, except those involving the target nucleus. Therefore, we can let couplings evolve through many gates and need to arrive at the goal coupling only when a noncommuting pulse affects either member of the coupling pair. The refocusing scheme can be made more efficient by the use of “virtual 180s.” Conventionally, every time a refocusing pulse is used, a second refocusing pulse must be applied at the end of the period to cancel the first one and ensure that all the spins return to the initial state, so that the computation is unaffected by the refocusing. While the first refocusing pulse is needed to physically refocus the coupling, the second one can be made virtual and not physically applied. The virtual 180 can be implemented by pushing it forward or backward (the virtual 180 can be created before or after the physically applied refocusing pulse) through the pulse sequence until it can be merged with another pulse. If it is another refocusing pulse, then it can cancel with the virtual 180 created at that refocusing event. Or, if it is a computational pulse, the virtual 180 can be absorbed by modifying the phase of the pulse and introducing a frame change (see Appendix C for details).

All of these techniques are combined into an efficient pulse compiler. The pulse compiler loads the information about the internal Hamiltonian, the effective pulse decompositions [Eq. (13)], and the desired pulse sequence with a suitable refocusing scheme. It can then optimize the delays between the pulses, the timing of the refocusing pulses, and the phases of the pulses to best implement the desired sequence. At any point in the sequence, phase or coupling error terms from the decomposition are compensated for by use of the phase-tracking calculations or are absorbed in the refocusing scheme (see Fig. 3). This allows for long pulse sequences to be constructed while avoiding excessive error accumulation. For further details of the sequence compiler, see Appendix D.

The sequence compiler can optimize only the delays between events and it must be given a suitable sequence of refocusing pulses to start with. Designing an exact refocusing scheme may require many refocusing pulses, and each refocusing pulse takes a finite duration and introduces its own errors. In some cases, the theoretical control accuracy gained by bringing the calculated ZZ coupling evolution closer to the goal is lost due to decoherence and pulse imperfections. There is therefore a trade-off between the theoretical accuracy of the control scheme and its duration and number of refocusing pulses. In practice this entails designing a good refocusing scheme and optimizing it. The total error from unfocused ZZ couplings is then calculated and judged whether it is acceptably low. If not, additional refocusing pulses are added to correct the errors and improve the optimization. In addition, a penalty function prevents the optimization from using excessive time for refocusing schemes.

IV. MERGING BOTH METHODS

Each method described above has its strengths and limitations. Optimal control theory can give robust and time-optimal control sequences for strongly coupled systems, where conventional pulse design fails. However, the method is intrinsically unscalable and is limited in the number of spins it can handle in practice. The number of qubits currently available in NMR already pushes these methods to their limit. The pre-error and post-error method with pairwise simulations provides a scalable, efficient solution to the design of control sequences but is limited by the properties of the predefined pulses, particularly the fidelity of the error-term representation. On long pulses targeting spins with strong couplings, the decomposition of Eq. (13) may fail to give a high-fidelity representation of the pulse, and better pulse engineering is needed. Here we show how the two methods can be combined to allow optimal control techniques to be applied to larger systems. A more computer science approach to scaling up optimal control techniques also utilizing subsystems has also been considered in Ref. [47].

The idea is to consider only a subsystem of the QIP’s qubits in designing the optimal control pulse. In NMR, particularly relevant subsystems consist of the spins of the same nuclear type, for example, all protons. A pulse that is designed on the subsystem without consideration for the other spins does not implement the desired unitary on the whole system. To determine the effect of the pulse on the entire system, pairwise simulations between the subsystem and the other spins are performed. These simulations capture and track both the evolution of the internal Hamiltonian on the other spins, and deviations due to couplings between the subsystems. The pulse on the full system can then be represented in the same manner as described above by adding pre-error and post-error terms, which can be accounted for during the optimization of the refocusing scheme and the phase-tracking calculations as part of a larger sequence.

There is no guarantee that the pulse decomposition with the error terms will give a high-fidelity representation. The
optimal control pulse drives the subsystem through some complicated trajectory, and it may not be possible to extract the effect of the couplings to other subsystems as simple ZZ errors before and after the ideal gate. In general this decomposition works well for short one-qubit unitaries but breaks down for two-qubit gates taking more time. However, whereas the pulse is optimized considering only one subsystem, it can be made robust to the effect of couplings to other subsystems by incoherently averaging over the states of the other subsystems as described in Ref. [48]. This is equivalent to making the pulse robust against Larmor frequency variations. Thus, averaging over all possible states is not necessary, and making the pulse robust to frequency shifts from the sum of the $J$ couplings is sufficient. This will make the pulse more difficult to find, but the obtained pulse will have a higher-fidelity representation on the full system.

To illustrate the basic ideas, consider a register of qubits organized into subsystem blocks, as illustrated in Fig. 4. Take, for example, the first subsystem of Fig. 4 consisting of three proton spins, where the other subsystems consist of other nuclear types. It may be that to effect a one-qubit rotation on qubit $Q_1$ by a simple method requires a very long selective pulse, because the chemical shift difference to the nearest spin is small. Then, the effect of the couplings cannot be taken into account by only $Z$ and ZZ pre-error and post-error terms, because $\Delta t J$ is too large. However, one can find an optimal control pulse that considers only subsystem 1 and that refocuses the intrasubsystem couplings while implementing the ideal gate. Perturbations due to intersubsystem couplings and the internal evolution of the other subsystems during the pulse can then be taken into account by pre-error and post-error terms.

The utility of such a scheme can be demonstrated on the crotonic acid molecule (see Appendix A for details of the molecule), where the natural subsystems are the three proton qubits and the four carbon qubits. Because of their refocusing properties, 180° rotations are more difficult to find and to represent. However, 180° rotations can easily be found for all the spins by use of this subsystem technique and then represented through subsystem pairwise simulations. For example a pulse on the methyl qubit is found by considering only the proton subsystem, and then, to determine the dynamics on the full system, simulations are performed on individual and pairs of the five subsystems, \( \{M, H_1, H_2\}, \{C_1\}, \{C_2\}, \{C_3\}, \text{ and } \{C_4\} \). The results are summarized in Table II and compared with the pulse representations of standard pulse designs. The crotonic acid molecule is not ideally suited to this approach because of the large couplings between subsystems; nevertheless, the subsystem GRAPE pulses have similar or better fidelities than standard pulses of a similar or longer length. In particular the GRAPE pulses have consistently high fidelities even where other pulses, such as selective pulses on $H_1$ and $H_2$, break down. Furthermore, since the clock of decoherence is always ticking, shorter pulses usually perform better.

Finding an optimal control pulse on the entire system such as in Fig. 1 may give shorter, higher-fidelity pulses than the subsystem with the error-term strategy described above. This is because the optimal control method is able to exploit more control handles. However, optimization over the full system is too difficult for large systems, whereas the methods used to find the pulses in Table II are scalable if the subsystems used are kept small. The combination of optimal control and use of pre-error and post-error terms is therefore

FIG. 4. Example of a block architecture: For clarity, couplings between qubits of blocks have been contracted into single arrows. It is assumed that the longer range couplings, such as those between subsystems 1 and 4, vanish. For example, in crotonic acid, subsystem 1 could be the proton spins, so that $Q_1, Q_2, Q_3$ are $M, H_1, H_2$, and subsystem 3 could be the carbon spins and there would be no further subsystems. In the inset an example of a subsystem optimal control pulse, in this case the pulse implements a 180° rotation on the methyl qubit of the crotonic acid molecule, is found considering only subsystem 1. The effects of the other subsystems are then characterized using a set of “subsystem” pairwise simulations. As previously, such simulations can be used to compute optimal pre-error and post-error terms that enable control of the couplings by optimization of the refocusing scheme and the phase-tracking calculations described in Sec. III B.
TABLE II. Comparison of the performance of optimal control pulses constructed and corrected by using the subsystem approach (last column) to more conventional shaped pulses (first two columns). The pulse lengths are given in $\mu$s and the percent fidelities are the Hilbert-Schmidt norm of the pulse including pre-error and post-error terms compared to the ideal pulse [see Eq. (13)]. All pulses implement one-qubit 180° rotations (see text). The crotonic acid molecule is not ideally suited to the subsystem approach because of the large couplings between subsystems. This required the effects of the other subsystems to be suppressed through incoherent averaging over chemical shifts of the $J$ coupling strength (see text and Ref. [48]) in order to obtain high fidelities.

|   | isech | Hermite 180 | GRAPE |
|---|-------|-------------|--------|
|   | Length | Fidelity    | Length | Fidelity | Length | Fidelity |
| $M$ | 896   | 99.50       | 2000   | 99.96    | 750    | 99.92    |
| $H_1$ | 3300  | 91.71       | 7000   | 97.71    | 3000   | 99.79    |
| $H_2$ | 3300  | 97.42       | 7000   | 97.42    | 3000   | 99.79    |
| $C_1$ | 128   | 99.92       | 300    | 99.97    | 60     | 99.95    |
| $C_2$ | 700   | 99.45       | 1400   | 99.93    | 700    | 99.92    |
| $C_3$ | 700   | 99.39       | 1400   | 99.94    | 700    | 99.85    |
| $C_4$ | 700   | 99.86       | 1400   | 99.96    | 700    | 99.88    |

well suited for designing a control sequence for the class of QIPs made of subsystems with strong internal couplings but weaker coupling between subsystems.

V. CONCLUSION

Liquid-state NMR has led to the development of numerical methods to find accurate pulses to control simple systems robust against errors from perturbations in fields, amplitudes, and other spins. Together with methods for accounting for or refocusing phases and couplings, we have a basis to efficiently design robust control sequences for QIPs made of weakly coupled blocks of qubits. Within blocks, faster dynamics are handled by pulses obtained by optimal control methods. Pre-error and post-error term analysis can be performed on such pulses to evaluate perturbations due to the other blocks. One can then combine such pulses to efficiently construct a sequence to implement the desired algorithm. Phase tracking and numerical optimization of the refocusing scheme allows us to correct pulse errors, thus avoiding error accumulation, and to implement gates between different subsystems. The combined model of control presented here can be applied to other Ising-type models. A potential example is in superconducting qubits, where the Hamiltonian can be written in a pseudospin form that is similar to the liquid-state NMR Hamiltonian [35]. It is conceivable that the control fields would affect only a small number of qubits at a time, thus ensuring the presence of natural subsystems to use for optimal control.

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APPENDIX A: CROTONIC ACID

Crotonic acid offers a total of seven qubits: Two protons and the methyl group and four $^{13}$C-labeled carbons. The methyl group consists of three magnetically equivalent protons that form spin-$\frac{1}{2}$ and spin-$\frac{3}{2}$ subspaces. With an appropriate pulse sequence, the spin-$\frac{1}{2}$ subspace can be exclusively selected and treated as a qubit. The internal Hamiltonian of the system is shown in Fig. 5.

APPENDIX B: GRAPE PULSE FINDER

The GRAPE code developed at IQC is closely based on the search algorithm presented in Ref. [41]. Here we empha-
size only the specific modifications we have implemented that help to achieve high experimental fidelity. The parameters for the initial random guess have a significant impact on the success of the search and the type of pulse found. For good experimental implementation, smoothly varying low-power pulses are best. Therefore, the algorithm is initialized with a smooth low-power guess. For example, in a 250-point pulse every 25th point is random with a maximum range of ±5% of the maximum power available, and cubic splines are used to interpolate the rest of the points. For finding unitary gates, the fitness function is the HS fidelity defined in Eq. (8). For good experimental fidelity the pulse is made robust to variation in rf field strength and/or calibration errors by taking an average of the fitness function defined over a range of rf field strengths. A three-point distribution at rf power multipliers of 0.95, 1, and 1.05 gives good results. Robustness to variation in internal Hamiltonian parameters such as chemical shifts can also be put into the fitness function; however, for typical liquid-state linewidths the pulses found are more than sufficiently robust. This feature can also be used for incoherent averaging in the subsystem approach. The convergence of the search algorithm is dramatically improved by the use of conjugate gradient techniques.

To minimize decoherence, it is desirable to minimize the pulse duration. However, because we do not a priori know even approximately what this duration should be, it would take many optimizations for different durations to find the optimum. Furthermore, the time optimal point may require unrealistically high powers or fast pulse variations. Instead, we allow the length of time for each step of the pulse to vary and use time-step derivatives to converge quickly to a good pulse duration. A penalty function is added to limit the total time of the pulse and push the optimization toward shorter pulse times.

For high-fidelity experimental control, smooth and slowly varying “spectrometer friendly” pulses are ideal, although the desired criteria can be enforced in the search with more sophisticated techniques [49]. These techniques require defining the pulse by many points. We have found a simple method to systematically find sufficiently smooth pulses for large systems, where defining pulses by many points is not feasible:

1. Find a high-fidelity pulse using relatively long time steps, e.g., 20 μs.

2. Digitally smooth this pulse with shorter time steps of, e.g., 1 μs, making sure that the pulse bandwidth is within the system limitations.

3. Use this smoothed version of the pulse as an initial guess for the numerical optimization. There will be a small loss of fidelity from the smoothing. Nevertheless, it will be a good starting point.

4. If necessary, repeat the smoothing and optimization procedure.

Empirically we have found that the reoptimization of the pulse in the third step changes the pulse very little, and once the new optimized pulse is determined, it is still sufficiently smooth. It is also necessary to ensure that the beginning and end of the pulse go to zero to ensure a smooth experimental turn on or turn off. This can be achieved with a penalty function that penalizes high powers at the beginning and the end.

In combining the two methods of control described in this paper we have discovered an additional method that can be used to reduce the time needed to find pulses on larger systems. This method provides a practical way to find certain types of pulses on large systems, but it does not address scalability issues. If the system can be decomposed into possibly overlapping subsystems such that for each subsystem, the desired unitary operator factors into one acting on the subsystem and another acting on the complement, then we can find good pulses by defining the fitness function as a weighted sum of fitness functions for each subsystem. The subsystems must be defined so that the dominant dynamics of the system is captured. In particular, every strong coupling must be internal to at least one of the subsystems. We simulate each subsystem separately to reduce the simulation complexity. Unlike the scalable subsystem approach discussed in Sec. IV, the optimal control subsystems will cover the entire system and the pulse is expected to implement the desired unitary on the full system without pre-error or post-error terms. This approach shares some similarities with recent work simulating larger spin systems with a subsystem-style approach [50].

Take as an example the crotonic acid system with seven qubits. The molecule is shown in Fig. 5. Pulses implementing one-qubit rotations have the desired factorization property with respect to any subsystem. To capture the strong couplings, we use the following four overlapping subsystems: \{M, C_1\}, \{C_1, C_2\}, \{H_1, C_2, C_3, H_2\}, and \{C_3, C_4\}. \(C_1\) is covered by two subsystems capturing its strongest couplings. If the pulse refocuses these strong couplings to \(M\) and \(C_2\) it is likely to also refocus weaker couplings to say \(C_4\) without additional optimization. In general though, if the pulse performs the desired subsystem unitaries with high fidelity there is no guarantee that the smaller unconsidered couplings and/or many-body effects will not give a much lower fidelity implementation on the full system. Empirically however, we have found the constructed pulses work well on the full system, both for short one-qubit gates and for longer coupling gates. Furthermore, it substantially speeds up finding pulses for the seven-qubit system. The pulse shown in Fig. 1 was found with this method. It has 99.9% fidelity averaged over the above subsystems and showed a 99.7% fidelity on the entire seven-qubit system.

**APPENDIX C: PULSE COMPILER**

The efficient pre-error and post-error term analysis described in Sec. III B contains two steps: The first is to efficiently compute the relevant dynamics of the system under the rf pulse; the second is to extract optimal pre-error and post-error terms that are correctable and give a good representation of the simulated pulse.

**Capture the relevant dynamics.** Consider an illustrative three spin system. Spins 1 and 2 are of the same species, whereas spin 3 is different. The propagator \(U\) of a pulse affecting spins 1 and 2 can be written as

\[
U = \exp(-i\Delta(t_{\text{int}}[\cosθ(X_1 + X_2) + sinθ(Y_1 + Y_2)]) + \omega_1Z_1 + \omega_2Z_2 + \omega_3Z_3 + J_{12}Z_1Z_2 + J_{23}Z_2Z_3 + J_{13}Z_1Z_3),
\]

(C1)

where the control and internal Hamiltonian terms are defined.
as above. If the control fields are functions of time, the pulse must be discretized and integrated over many time steps.

The dynamics of the system can be expanded into contributions of different coupling order. If coupling effects are ignored, then the zeroth order $U_0$ can be computed by three independent single-spin simulations as

$$U_0 = e^{-i\Delta t [\omega_{1} \cos(\phi_1 + \phi_2) + \omega_{2} Z_1]} \times e^{-i\Delta t [\omega_{1} \cos(\phi_1 + \phi_2) + \omega_{2} Z_2]} e^{-i\Delta t (\omega_{3} Z_3)}. \quad (C2)$$

The effects of the coupling can then be added in a perturbative manner. The first step is to consider the effect of the couplings between pairs of spins or subsystems: couplings 1-2, 2-3, and 1-3 are considered. By considering only pairs of spins, coherent indirect coupling effects are lost. For example, there is an extra effect on the dynamics of qubit 1 from the coupling to qubit 2 because qubit 2 is in turn affected by the coupling to qubit 3. These higher-order coupling effects can be neglected when coupling effects during the pulse are small. Moreover, for soft selective pulses in the weak-coupling approximation, many of the second-order contributions vanish because all the internal Hamiltonians of the off-resonant spins commute with each other, so that in most cases, stopping at first order is sufficient to encompass all the relevant dynamics.

If the pulse is found using optimal control theory rather than a predefined simple pulse shape, it may drive the system through a nonintuitive path that makes use of indirect coupling dynamics to achieve the desired unitary. Therefore we need to perform a set of simulations between the subsystem defined for the optimal control and each of the other spins. For example, a GRAPE pulse may be found by considering only the carbon subsystem. Each pairwise simulation encompasses the carbon spins and one of the proton spins. The computational resources required are larger than for a simple spin pairwise simulation, but they still scale polynomially with the total register size, provided the subsystems are of constant size.

**Search for the optimal error terms.** From the efficient simulations of the preceding paragraph, we need to determine the pre-error and post-error terms surrounding the ideal operation. This is achieved through a simple numerical search optimizing the following fitness function:

$$\Phi = \text{Re}[\text{tr}(U_{\text{sim}}^\dagger (U_{\text{pos}} U_{\text{ideal}} U_{\text{pre}}))] / N. \quad (C3)$$

This compares the simulated unitary to its representation as the ideal unitary, restricted to the relevant qubits, sandwiched between the pre-error and post-error terms. This fitness function is sensitive to global phase because we are considering only a subsystem, and so to be consistent when representing the gates on the full system, local phases matter. To extract the optimal error terms one can use a classical search algorithm. Note that this search is performed on a relatively low-dimensional parameter space. For pulses designed for one qubit, the single-spin simulations require only two parameters for the phases in the error terms on this spin. In the pairwise simulations there are six parameters for the phases in the $Z$ and $ZZ$ error terms on the coupled spins.

**Error-term computation.** For standard shaped pulses designed to act independently on one or more spins, the following procedure allows one to compute error terms for the pulse representation.

1. Perform single-spin simulations including the rf control fields for each spin and optimize the $Z$ error terms (with respect to the intended evolution on the spin) to capture the effects of the chemical shift and the transient Bloch-Siegert effect.

2. Perform simulations of each pair of spins, including the pairwise coupling and the rf control fields. Optimize the $Z$ and $ZZ$ error terms with respect to the intended evolution on both spins. Remove the contribution of the $Z$ error terms found in the first step from the ones found here (by dividing the terms or subtracting the exponents) to account for the fact that the simulation also accounts for the effects modeled in the first step. Call these the “incremental $Z$ error terms.”

3. Determine the $Z$ error terms for each spin as the product of the terms found in the first step and the incremental terms from each pairwise simulation involving this spin. The $ZZ$ error terms are the ones found in the second step.

If the pulse is designed to have an independent action on a set of subsystems, the procedure needs to be modified by replacing the single-spin and pairwise simulations with single-subsystem and pairs-of-subsystems simulations. This requires optimizing both $Z$ and $ZZ$ error terms for the single-subsystem simulations, and determining incremental contributions for the pair simulations by removing the contributions already obtained for spins and pairs internal to each subsystem.

**APPENDIX D: SEQUENCE OPTIMIZER**

The sequence compiler was built to automate, as much as possible, the design of pulse sequences for an NMR QIP. The compiler takes as one input the fixed information about the internal Hamiltonian, namely the chemical shifts and $J$ coupling values, and the error-term information for the pulses obtained by optimal control and implementing the set of one-qubit rotations needed. The second input is a representation of a goal sequence of quantum evolutions that implements the desired algorithm or quantum network. The sequence compiler then determines delays between pulses so as to minimize the total error of the implementation compared to the intended evolution. The goal sequence is described with a purpose-built language. Here is an example of such a sequence:

```plaintext
:pulse C190 0 @C1:X+
:zz 0.25 C1 C2
:refocus C3180 0.25
:pulse C290 0.75 @C2:0+
;z 0.5 C3
```

This sequence requests a $90^\circ$ rotation about the $x$ axis on $C_1$ followed by a ZZ$90^\circ$ (equivalent to a controlled-$Z$ gate up to single-qubit $Z$ rotations) coupling gate between $C_1$ and $C_2$ followed by a $90^\circ$ one-qubit gate on $C_2$, and a final $Z$ rotation on $C_3$. During the coupling period a “floating” refocusing pulse of $180^\circ$ about the $y$ axis is executed on $C_3$. The @ instructions specify state assumptions that may simplify the optimization. The output of the compiler is a pulse sequence.
that can be directly executed by the spectrometer. The compiler automates a number of the tasks discussed in Sec. III B.

(i) Phase tracking. To avoid having to use one physical spectrometer channel per qubit, the compiler tracks the evolution of the nuclear rotating frames and the spectrometer channel frames throughout the computation. When a pulse about a certain axis is required for a nucleus, a simple calculation of the phase difference between the channel frame and the nuclear frame determines the phase at which the pulse should be sent in the channel frame to achieve a specific phase in the target nuclear frame. As noted above this freedom allows us to avoid having to refocus the chemical shifts at every step. It also allows free (pulse-less) implementation of $Z$ rotations by executing a frame change on the target nuclei and updating the phases of all subsequent pulses. In addition, the relative phase evolution of the nuclei and the observation channel are tracked and the observation phase is adjusted appropriately. The frequency changes can be implemented by either changing the transmitter frequency or phase ramping the pulse. The phase-tracking calculations assume that the frequency change is phase coherent, but experimentally this is not always the case (particularly for large frequency changes), and we have achieved more consistent results with phase ramping on our spectrometer.

(ii) State assumptions. In some algorithms the state of the system might be known at certain steps. For example, at the beginning of the algorithm we may know that a particular qubit is in the state $|0\rangle$ or the maximally mixed state $I$. This can simplify the refocusing and phase-tracking calculations. For example, if two spins are in the maximally mixed state, the coupling between them has no effect and so does not need to be refocused. If a qubit is in a pseudopure state then its coupling effect with the other spin is reduced to an additional $Z$ rotation, which can be accounted for with the phase-tracking calculation. State assumptions are implemented by specifying the nuclear states after pulse commands, as shown in the example. The compiler also implements some elementary state update rules.

(iii) Cancellation of virtual 180s. A refocusing pulse refocuses the couplings as desired but also affects the computation. As discussed in the text, conventionally this has been avoided by placing refocusing pulses in pairs to ensure that the state of the refocused spin is not changed. However, this second refocusing pulse can be considered a virtual 180 that does not have to be implemented physically. As such it can be implemented by absorbing it into the next operation, modifying this operation appropriately. Using the notation $R_{\phi}(\theta)$ to denote a rotation of $\theta$ about the axis an angle $\phi$ away from the $x$ axis in the $x$-$y$ plane, the following rotations are equivalent:

\[
R_{\phi}(\pi) = R_{\phi}(\pi) = \gamma R_{\phi}(\frac{\pi}{2})
\]

where we use matrix multiplication and trigonometric identities to determine $\gamma$ and $\delta$,

\[
\gamma = 2(\alpha - \beta),
\]

\[
\delta = 2\beta - \alpha - \pi.
\]

That is, a $180^\circ$ pulse at phase $\beta$ followed by a $90^\circ$ pulse at phase $\alpha$ is equivalent to a $90^\circ$ pulse at phase $2\beta - \alpha - \pi$ followed by a rotation about the $z$ axis of $2(\alpha - \beta)$. Since the $z$ rotation comes for free as a frame change, we have eliminated the need to do the second 180. The compiler keeps track of these virtual 180s, which considerably simplifies writing pulse programs. The virtual 180 can also be sent backward through the pulse program in a similar manner and absorbed into a preceding pulse. In some cases this may help the refocusing scheme.

(iv) Optimization of delays. The pulse sequence can be considered as a sequence of events (computation and refocusing pulses) with delays in between. The delays between pulses serve the dual purposes of allowing desired coupling gates to occur and also allowing time for refocusing unwanted coupling effects. The refocusing pulses change the direction in which the couplings are evolving, which can help reach the coupling goals. At the beginning of each event certain couplings must be at their goal values. In most cases, only those couplings that do not commute with the pulse’s intended effect have a fixed goal target. Other couplings are simply tracked until a fixed goal is required [46]. The coupling evolution for each pair of spins is calculated at each event from both the coupling evolution during the delays and the coupling error terms in the pulse representations. For those pairs that have a fixed target at this event the calculated coupling is compared to the goal. A Euclidean distance function is defined as the sum-squared error between the goal and the coupling is compared to the goal. A Euclidean distance function is defined as the sum-squared error between the goal and actual couplings and is related to an estimate of fidelity loss for the whole sequence. Optimizing the pulse sequence is now reduced to the task of optimizing the delays between each period. This is handled by a simple iterative optimization to minimize the total contribution to the distance function of the events bounding the delays. That is, the delays are individually optimized one by one, starting at the last one. After the first one is optimized, the sequential optimization starts again at the last one, repeating the process until the improvement is smaller than a threshold, or a goal distance is achieved. Although effective, this optimization strategy is simplistic and easily gets trapped in local minima. It would be useful to develop strategies that optimize all delays together with a nonlinear least-squares optimization and consider different distance functions such as maximal error.
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