A path towards distributed quantum annealing

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Quantum Annealing has proven to be a powerful tool to tackle several optimization problems. However, its performances are severely limited by the number of qubits we can fit on a single chip and their local connectivity. In order to address these problems, in this work, we propose a protocol to perform distributed quantum annealing. Our approach relies on Trotterization to slice the adiabatic evolution into local and non-local steps, the latter which are distributed using entanglement-assisted local operations and classical communications (eLOCC). Theoretical bounds on the Trotter step size and successful distribution probability of the process have been established, even in the presence of noise. These bounds have been validated by simulating numerically the evolution of the system, for a range of annealing problems of increasing complexity.

INTRODUCTION

Solving optimization problems is a very common problem in science with many practical applications, making this a very active field of research. [1][3]. For these problems physics has contributed with lots of concepts and methods to the field of optimization, starting from the idea of thermal simulated annealing [4], to the applications of replica and cavity methods [3]. Quantum Annealing was initially formulated in the 90’s [5] as a quantum alternative to classical simulated annealing in which quantum tunneling replaces thermal hopping in order for the system to avoid being trapped in local minima and reach the ground state (i.e. the solution of the optimization problem). This idea was later developed to the point at which, today, we have working quantum annealers with thousands of qubits available.

However, Quantum Annealers suffer from some issues that are limiting the size and complexity of the problems we are currently able to solve using this technology. The major problems that we will be addressing here are embedding and connectivity, which are closely related. The connectivity problem is simply an engineering bounds to the number of qubits that one is effectively able to couple on the chip layout (in the latest generation quantum annealers the number of couplings per qubit is \( \sim 15 \)). This limits the problems that one can solve on a quantum annealer to ones that are sparsely connected. A way to circumvent this problem is to use minor-embedding techniques in which a chain of physical qubits is treated as one logical qubit allowing for more connections. This technique, however, is pretty intense in the number of qubit used thus limiting the size of the problems which we can effectively solve.

Recently, advances in the distribution of quantum operations [6][3] opened the possibility of connecting distant qubits and perform operations at a distance. This has led to the emergence of new paradigms such as distributed quantum computation [9][12], by which multiple quantum computers can be effectively entangled to exploit their joint computational power. Distribution is an interesting prospect for Quantum Annealers, as it could improve their scalability and connectivity. However, so far, no protocol has been proposed for Quantum Annealing, neither it has been investigated the possibility of expanding these techniques to this particular model of computation.

After a brief formal introduction of Quantum Annealing, in this work we will propose a protocol for the distribution of quantum annealing problems across multiple machines. We will also investigate the robustness of the protocol in the face of imperfect distribution and perform numerical simulations on several problems.

BACKGROUND

Quantum annealers are a special type of analog quantum computers, making their inherent process continuous in time. It is governed by a time-dependent Hamiltonian \( H(t) \).

\[
H(t) = \left(1 - \frac{t}{t_F} \right) H_0 + \frac{t}{t_F} H_F. \tag{1}
\]

At first, the system Hamiltonian is \( H_0 \), which acts on the initial state \( |\psi_0\rangle \), the ground state of \( H_0 \). While the system Hamiltonian slowly evolves to \( H_F \), its state evolves to \( |\psi_F\rangle \). According to the adiabatic theorem (see [13] for a thorough review on adiabatic quantum computation), the higher the annealing time \( t_F \) is, the closer the final state \( |\psi_F\rangle \) will be to the ground state of the final Hamiltonian \( H_F \).

To model the system of qubits we will consider the Transverse field Ising Model, and so the starting and final Hamiltonians \( H_0 \) and \( H_F \) take the form in eqs. [3] and [5], respectively. The qubits are labeled by the ordered set \( \Omega = (q_0, q_1, ..., q_{N-1}) \). The \( h_i \) terms correspond to local fields applied to the qubit \( i \), which are represented by nodes in a graph fig. [4]. With 2-body Hamiltonians, we have the couplings \( J_{ij} \) which quantify the strength of the \( \sigma^z \sigma^z \) interactions between the
qubits $i, j$. The qubits in the set $\Omega$ then are connected by the edges $e_0$ that indicate which qubits are coupled, i.e. $e_0 = \{(i, j) : J_{ij} \neq 0\}$.

$$H_0 = \sum_{i \in \Omega} \sigma_i^x$$

(2)

$$H_F = \sum_{i \in \Omega} h_i \sigma_i^z + \sum_{(i,j) \in e_0} J_{ij} \sigma_i^z \sigma_j^z$$

(3)

In the context of computation, this process is useful as an optimization meta-heuristic, where the final Hamiltonian $H_F$ encodes the objective function. Furthermore, having $H_F$ expressed in the Ising Model, quantum annealing becomes an ideal process for solving Quadratic Unconstrained Binary Optimization (QUBO) problems \[14, 15\]. These problems, other than being generally hard to solve, are often defined on graphs with a large number of connections. This limits the use of current generation quantum annealers as they suffer from connectivity problems as discussed above.

Distributed quantum computation allows for the use of smaller quantum computers, with more manageable noise and therefore could obtain better qubit coherence. Arbitrary quantum operations between two distant parties can be achieved through having a classical and a quantum channel between both parties \[16\]. This channel can be made, for example, by entangled photons in fibers \[17\] or by having a superconducting waveguide cryogenic “cable” connecting them \[6\]. Another of the main challenges of distributed quantum computation is to have an efficient and low noise interface between the entanglement carriers and the computing qubits. For example, optomechanical transducers are actively being research for being a promising technology that could allow long distance communication for superconducting technology \[7, 18–26\]. The last step, is to put all of this together to implement a distributed operation on two qubits. Experimentally, the state-of-the-art distributed CNOT operation has achieved a fidelity of $\sim 85\%$ for trapped ions \[8\].

It is then clear to imagine how distributed quantum computation techniques applied to quantum annealing can allow for virtually unlimited connectivity between qubits allowed to interact non-locally (we will call them the interface qubits). Moreover, having the qubit connectivity not restricted to a fixed hardware topology will help solve the connectivity problem present in current quantum annealing devices. Another advantage of distribution is that it allows to take advantage of smaller quantum devices, which are easier to manufacture and manage, therefore improving the scalability. In the next section we describe the theoretical details of the proposed distribution approach before moving to simulations.

DISTRIBUTED UNITARIES FOR QUANTUM ANNEALING

To formulate the proposed distributed quantum annealing protocol, we start by making a distinction between local and non-local Hamiltonians. Local Hamiltonians act on a single annealing processor, while non-local ones are only responsible for coupling qubits which belong in spatially separate processors (see fig. 1). In general, any evolution with time-dependent Hamiltonians can be separated into local $H_L(t)$ and non-local $H_N(t)$ Hamiltonians.

![Figure 1](image.png)

**FIG. 1.** Example of a configuration for a distributed quantum annealing network. The qubits native to the quantum annealers and their couplings are colored black. The blue areas denote separate annealing systems, which rely on the local, analog evolution of the separate quantum annealers. In this case, $H_L = H_{LA} + H_{LB}$. The red dashed lines represent the non-local couplings, belonging to the non-local Hamiltonian $H_N$. The red points inside the qubit nodes indicate that those qubits are interface nodes, therefore being able to communicate non-locally. The interface qubits can communicate to any other interface qubit reachable by an entanglement distribution network, which immensely increases their connectivity relative to local qubits.

$$H(t) = H_L(t) + H_N(t)$$

(4)

The evolution process which happens in quantum annealing can be expressed as a unitary operator, with the use of a time-ordered exponential \[6\]. The process can then be decomposed into a sequence of $M$ unitary operators \[6\].

$$U(0, t_F) = T_t \exp\left\{-i \int_0^{t_F} dt H(t) \right\}$$

(5)

$$= T_{t_k} \prod_{k=0}^{M-1} U(t_k, t_{k+1})$$

(6)

where $T_t$ is the time-ordering operator, $t_k = k \Delta t_k$ and $\Delta t_k = t_F / M$ is the Trotter step size (chosen to be constant for simplicity). Then, each $U(t_k, t_{k+1})$ is Trotterized into the local and non-local parts

$$U(t_k, t_{k+1}) \rightarrow U_N(t_k, t_{k+1}) U_L(t_k, t_{k+1}).$$

(7)
So that we get the whole Trotterized evolution

\[ \tilde{U}(0, t_F) = T_k \prod_{k=0}^{M-1} U_N(t_k, t_{k+1}) U_L(t_k, t_{k+1}). \] (8)

The local unitary evolution operators \( U_L(t_k, t_{k+1}) \) represent a partial evolution of the quantum state inside the quantum annealing processor. These alternate with the non-local unitary operators \( U_N(t_k, t_{k+1}) \), which simulate and distribute the non-local couplings between qubits in separate quantum processors. These are implemented via a unitary distribution protocol, i.e. telegates, such as the one proposed in [10]. Therefore, the distributed annealing process corresponds to the fast switching between local Hamiltonians, the latter which are implemented via telegates. A similar process which Trotterizes between analog evolution and digital gates on a single quantum processor has been first proposed in [27] and realized experimentally in [28].

**Graph Split**

As pointed out before, annealing Hamiltonians can be represented with graphs, with vertices labeled \( \Omega = (q_0, q_1, ..., q_{N-1}) \) and edges \( e_0 \). The annealing graph then has to be split into multiple quantum annealers, making the identification of the local and non-local couplings. In this work we propose two ways to do so: edge splitting and vertex splitting. Consider two quantum annealers \( A \) and \( B \). An original annealing problem represented by \( G_0 \) then turns into local graphs \( G_A, G_B \) (where \( G_A + G_B \) has the terms in the local Hamiltonian \( H_L \) and non-local edges \( e_{AB} \). In edge splitting, some of the couplings in \( e_0 \) turn into non-local couplings \( e_{AB} \), such that \( G_A \) and \( G_B \) are disjoint. Vertex splitting is a slightly different. Instead of turning a (local) coupling into a non-local coupling, a vertex (qubit) is n-plicated across multiple quantum annealers. For instance, a qubit labeled \( q \) in the original problem can be duplicated to annealers \( A \) and \( B \), into qubits \( q_A \) and \( q_B \). This sounds suspiciously like quantum cloning [29], but it is not the case here. It relies on the initial Hamiltonian \( H_0 \) to be such that the state of the system evolves in a way that keeps the n-plicated qubits aligned. The mechanism is described in appendix A. For demonstration purposes, consider the following Hamiltonians representing an annealing problem:

\[
H_0 = \sigma_a^x + \sigma_b^x + \sigma_c^x,
\]

\[
H_F = J_{ab} \sigma_a^x \sigma_b^x + J_{ac} \sigma_a^x \sigma_c^x + J_{bc} \sigma_b^x \sigma_c^x.
\]

**Convergence of Trotterization**

The Trotterized evolution introduces errors due to \( H_L \) and \( H_N \) not commuting with each other. At each time interval \( t \in [t_k, t_{k+1}] \), there may be an equivalent Hamiltonian \( H_{eq}(t_k) \) that governs the process as if it was a continuous process. Assuming a uniform time-step, its existence is guaranteed when (see appendix B for proof)

\[
t_{k+1} - t_k = \Delta t < \Delta t_M
\] (9)
This result indicates that the equivalent Hamiltonian approximates the adiabatic Hamiltonian for small enough $\Delta t$. We conclude that, in ideal scenarios (no noise and no limitations for $\Delta t$ or $M$), the Trotterized process is equivalent to quantum annealing.

### Noisy Distribution

Ideal distribution is equivalent to quantum annealing, but it is not realistic. Therefore we will investigate the behaviour of our protocol in the case where, instead of distributing with pure Bell states, we consider a noisy quantum channel with some entanglement fidelity $F_\Phi < 1$. This analysis will also help us understand if this protocol is robust enough to noise to be implemented in a real-world scenario. The shared quantum state is

$$\rho_\Phi = x |\Phi\rangle \langle \Phi| + \frac{1-x}{4} I_{2 \times 2}$$

with fidelity

$$F_\Phi = x + \frac{1-x}{4}.$$

The CNOT distribution protocol (DCNOT) consumes a Bell state in the process

$$|\Phi\rangle = \frac{1}{\sqrt{2}} (|00\rangle + |11\rangle)$$

The fidelity of the state between the ideal distribution and the noisy distribution has the bound

$$p_N(\delta) > \left(1 - \frac{4}{3} \delta \right)^{M_D}$$

where $M_D$ is the number of consumed Bell pairs (same as the number of DCNOT gates, and is proportional to the number of Trotter steps $M$), and $\delta = 1 - F_\Phi$ is the infidelity of the shared entangled pair. This expression is deduced in appendix C. We have found this to be true for any distributed process that employs noisy telegates.

### NUMERICAL SIMULATIONS

In this section, we look at how the method performs with different annealing problems, starting with the spin chain, and later with other more densely connected networks. There are two important quantities that we will use to compare the Trotterized process (T, ideal or noisy distribution) with the annealing evolution (A) or the ground state space of $H_F$, corresponding to the limit when the annealing time $t_F \to \infty$. These are the fidelity $p_{*0}$ between the evolved state ($* \in \{A, T\}$) and the ground state space of $H_F$ and the relative energy error $\varepsilon_{*0}$ between $* \in \{A, T\}$ and $\bullet \in \{0, A\}$, in eqs. (14) and (15).

$$p_{*0} = \sum_{|\phi_i\rangle \in gs(H_F)} \langle \phi_i | \rho_* | \phi_i \rangle, \quad * \in \{A, T\}$$

where $gs(H_F)$ is the ground state space of the final Hamiltonian $H_F$.

$$\varepsilon_{*0} = E_* - E_{\text{mix}} \quad * \in \{A, T\}, \quad \bullet \in \{0, A\}$$

$E_*$ is the energy of the state after evolution, measured in the final Hamiltonian $H_F$, with $E_A$ or $E_T$ for adiabatic and Trotterized evolutions, respectively. $E_0 = \langle \phi_0 | H_F | \phi_0 \rangle$ is the ground state energy of $H_F$. $E_{\text{mix}} = \langle \phi_0 | H_F | \phi_0 \rangle$ is the energy of the initial state $|\phi_0\rangle$ measured in $H_F$. This corresponds to the average energy of a state after a process that randomly guesses the minimum of $H_F$, while respecting the evolution dynamics of $H_0$. It is, in principle, the worst case scenario.

With this definition of the renormalized energy error, we have that $\varepsilon_{*0} \sim 0$ when the evolution successfully finds the ground state (or one of the ground states, if $H_F$ is degenerate), and $\varepsilon_{*0} \sim 1$ when no useful information about the ground state can be obtained from $|\psi_*\rangle$. These qualities are also true for the probability $1 - p_{*0}$. There is also the case when the adiabatic evolution doesn’t find the ground state $\varepsilon_{A0} \sim 1$. In this case, the Trotterized process might still be able to approximate the adiabatic evolution in case $\varepsilon_{TA} \sim 0$.

#### Toy model

The first set of simulations will be about a simple graph network - the four qubit spin chain. The graph splitting for the spin chain is shown in fig. (a). There are 3
tunable parameters related to the Trotterization process:

- The annealing time $t_F$, which controls the speed of the adiabatic evolution;
- The number of Trotter steps $M$, or equivalently the Trotter step size $\Delta t = t_F/M$, which control the accuracy of the Trotterization;
- The noise of the distribution process, tuned via the entanglement fidelity $F_\Phi$.

The overall effect of each of the parameters can be seen in Fig. 4(a), indicated with different colored arrows. Firstly, a gradient is seen as the annealing time $t_F$ increases and the relative energy error $\varepsilon_{TA}$ decreases. This effect is due to the adiabatic nature of the process: the slower the process is, the closer the final state is to the ground state of $H_F$, and therefore $\varepsilon_{TA} \rightarrow 0$. Secondly, there is a clear transition region at constant $\Delta t$, which was expected from the theoretical analysis. This transition is visible at roughly $\Delta t \sim 1.0$, while the theoretical prediction is

$$\Delta t_M \equiv \min_k \left( \left| H_L(t_k) \right|^{-1} \left| H_N(t_k) \right|^{-1} \right) = 0.5,$$

being less than the one observed in the simulated Trotterization. The region where the algorithm appears to converge for $\Delta t > \Delta t_M$ is likely related to the conditional convergence of the equivalent Hamiltonian $H_{eq}$, where our theoretical analysis cannot establish any bound. Thirdly, a gradient is visible as the number of Trotter steps $M$ increases. Since $F_\Phi < 1$ there is distribution noise. With bigger $M$ there are more distribution steps, and naturally the errors due to distribution start being more noticeable.

Due to the adiabaticity condition (slow evolution, large $t_F$) the Trotterized adiabatic evolution is similar to the evolution of Floquet systems for intermediate time intervals $t \in [\tau, \tau + \Delta \tau]$, $\tau \in [0, t_F - \Delta \tau]$ with $\Delta t \ll \Delta \tau \ll t_F$. The phase transition we have observed is also a common feature of Floquet systems [31][34]. The region that remains at low energy, i.e. where $\varepsilon_{TA} \sim 0$, is the localized phase, and is associated with low ergodicity. In contrast, the region where $\varepsilon_{TA} \sim 1$ has energy close to the average of the eigenenergies of the Hamiltonian, and so is analogous to a system with infinite temperature, i.e. high ergodicity. Thus it is commonly referred to as the thermalization phase. In the literature this phase transition is known as the many-body localization (MBL) transition.

Fig. 4(b) shows the plot of the relative energy error between the Trotterized and the adiabatic evolution $\varepsilon_{TA}$, versus $\Delta t$. We can identify 3 regions. An initial one with $\Delta t < \Delta t_M$, where we are assured of the convergence of the equivalent Hamiltonian $H_{eq}$, and where we verify that $\varepsilon_{TA} < \varepsilon_{A0}$, meaning that the energy of the state after Trotterization $E_T$ is closer to the energy of the state after annealing $E_A$, than $E_A$ is to the ground state energy of $H_F$. It follows that in this region the energy error of the Trotterized evolution $\varepsilon_{TA}$ is almost all due to the annealing error $\varepsilon_{A0}$, and corresponds to the localized phase. Then, a region where $\Delta t > \Delta t_M$ and the error is still bounded, is where the phase transition occurs. We speculate that it is characterized by the conditional convergence of $H_{eq}$, which we have not been able to determine analytically. Finally, a third region where $\varepsilon_{TA} \sim 1$ is linked to high ergodicity and infinite temperature, and corresponds to the the divergence of $H_{eq}$. Here, the Trotterized evolution cannot be replaced by a continuous process governed by the Schrödinger equation, and as such the solution reached by distributed quantum annealing will not resemble quantum annealing.
Spin Chain and Sparse Networks

In this section we consider two different models that have more complex coupling networks. These are represented in fig. 5. Firstly, for all of the different network topology, we fixed the annealing energy error $\varepsilon_{A0} \sim 10^{-4}$ by tuning the annealing time $t_F$. Then for the ideal Trotterized evolution we have verified that with $\Delta t = \Delta t_M$, the relative energy error $\varepsilon_{TA} \lesssim \varepsilon_{A0}$ with ideal distribution ($\delta = 0$). From these simulations we know that the ideal Trotterization approximates well the evolution by annealing with $\Delta t = \Delta t_M$, indicating also that we are on the localized phase.

When we introduce noise, we have observed that the distribution fidelity eq. (13) is a crude lower bound. We introduce a free parameter $\beta$, and we observe that it tightens the bound considerably (see fig. 6 (a)). The distribution fidelity then becomes

$$p_N(\delta; \beta) = \left(1 - \frac{4}{3} \delta (1 - \beta)\right)^{MD}.$$ (16)

The motivation for the $\beta$ parameter is given in appendix C. The estimation in eq. (13) is too coarse as it neglects contributions to the probability of distributions with at least one distribution fault. Having a $\beta > 0$ indicates that the process can still have a few distribution faults, without hindering the fidelity evolved state with the ground state space of $H_F$.

It is plotted as a function of the entangled pair infidelity $\delta$, for some spin chain models labeled by $N, s$ (explanation of the model in fig. 3 (a)). The simulated data is shown in contiguous lines. The predicted curves where no distribution fault is allowed correspond to the dashed lines. The adjusted theory, with the $\beta$ parameter fitted to the data, is shown in dotted lines. These overlap for with the data for the majority of the values of $\delta$. This is not the case for larger $\delta$, where the observed log negativity plateaus. This is because $p_{TA}$ evaluated for the maximally mixed state is $1/2^{n(\Omega)} > 0$. This plot also shows that for small enough distribution error (small $\delta$), the Trotterization/annealing errors dominate, and thus the distribution error can be ignored. (b) The fit parameter $\beta$ for a few of the tested models, plotted against the total number of qubits in the graph $N$. The color of the points indicates the degeneracy of the ground state of the final Hamiltonians $H_F$ ($n(gs)$). For all of the models we have $\beta > 0$, which indicates that eq. (13) is a safe bound. The darker colors correspond to higher degeneracy and also larger $\beta$. This correlation can be understood as follows: the larger the ground state space of $H_F$, the more is a state affected by noise to be measured within it. Interestingly, the sparse networks, which have more non-local connections appear to be less sensitive to noise in each connection, when compared to the spin chain models with not many non-local couplings.

**Fig. 5.** Example of models with Spin Chain and Sparse Network graph topology. The coupling graphs have been parametrized in different ways to keep control on the number of qubits, non-local couplings and local systems. (a) **Spin Chain:** Parametrized by the number of qubits $N$ and the number of sublocal systems $s$. The shown example has $(N, s) = (6, 3)$. The local fields are $h_i = 0$, the local couplings are $J_{L} = 1$ and the non-local couplings $J_{N} = -2$. The number of non-local connections is tuned by changing $s$. (b) **Sparse Network:** Parametrized by the number of qubits $N$, number of qubits on system $0$, with $l < N$. The example graph has $N = 7$ total qubits, $l = 3$ qubits on sublocal system 0 and $N - l = 4$ qubits on sublocal system 1. The non-local connections are tuned by changing the number of interface qubits on either side. These are $i_0, i_1$ for interface qubits in sublocal system 0,1 respectively. Therefore, the example graph is parametrized by $(N, l, i_0, i_1) = (7, 3, 3, 2)$. The local fields are $h_i = 0$ and all couplings shown in the graph have $J = \pm 1$, with the sign choosen at random, given a generation seed.

**Fig. 6.** (a) Log negativity of the probability of measuring the Trotterized state at the ground state of $H_F$. It is plotted as a function of the entangled pair infidelity $\delta$, for some spin chain models labeled by $N, s$ (explanation of the model in fig. 3 (a)). The simulated data is shown in contiguous lines. The predicted curves where no distribution fault is allowed correspond to the dashed lines. The adjusted theory, with the $\beta$ parameter fitted to the data, is shown in dotted lines. These overlap for with the data for the majority of the values of $\delta$. This is not the case for larger $\delta$, where the observed log negativity plateaus. This is because $p_{TA}$ evaluated for the maximally mixed state is $1/2^{n(\Omega)} > 0$. This plot also shows that for small enough distribution error (small $\delta$), the Trotterization/annealing errors dominate, and thus the distribution error can be ignored. (b) The fit parameter $\beta$ for a few of the tested models, plotted against the total number of qubits in the graph $N$. The color of the points indicates the degeneracy of the ground state of the final Hamiltonians $H_F$ ($n(gs)$). For all of the models we have $\beta > 0$, which indicates that eq. (13) is a safe bound. The darker colors correspond to higher degeneracy and also larger $\beta$. This correlation can be understood as follows: the larger the ground state space of $H_F$, the more is a state affected by noise to be measured within it. Interestingly, the sparse networks, which have more non-local connections appear to be less sensitive to noise in each connection, when compared to the spin chain models with not many non-local couplings.
Comparison with standard annealing methods

So far, we have been concerned with the characterization of the process, increasing the complexity of the simulated models while doing so. In this section, we will directly compare the embedding tricks used in quantum annealing with the distributed alternative. We restrict the allowed topology of the annealing (local) hardware to a preset topology. We denote the original annealing problem by $A_0$, whose coupling graph is the fully connected 4 vertex graph. The hardware embedded equivalent is $A_1$, and has an additional parameter $J_M$ that corresponds to the strength of the negative couplings that are needed to keep the qubits in the same chain correlated. Then, there are two ways to distribute $A_0$: either by vertices ($D_0$), or by edges ($D_1$). Both require 4 non-local connections, but in $D_0$ they do not depend on the annealing problem, while in $D_1$ the strength of the non-local couplings is controlled by the annealing problem (encoded into $H_F$). These graphs are displayed in fig. 7.

While simulating the evolution of model $A_1$ we have observed that the system is harder to anneal (larger $\varepsilon_{A_0}$ for a given $t_F$) as $-J_M$ increases. This is likely due to a reduction of the minimum energy gap $\Delta\epsilon$, increasing the annealing time $t_F$. With a reasonable energy error $\varepsilon_{A_0} \approx 10^{-2}$, the required annealing time for the models $A_0, D_0$ and $D_1$ is $t_F \approx 4$, while for $A_1$ we need $t_F > 10$ (see fig. 5). We assess that for our tested models, the annealing curves ($\varepsilon_{A_0}$ vs $t_F$) of the distributed models $D_i$ are equivalent to the annealing curve of the original model $A_0$. The reason for this is that, for the edge splitting model $D_1$, its annealing Hamiltonian ($H = H_L + H_N$) is precisely the same as the original problem $A_0$. This is not the case for the vertex duplication model $D_0$, although we see an exact overlap of the annealing curves. Because of this we are confident that, as long as no qubit chains are employed, the distributed annealing problems have the same annealing curves as the original problem. From here, we then conclude that distributed quantum annealing without noise has the same time scaling as annealing with no embedding. This means that the distributed protocol offers no speedup when compared to a quantum annealer that can already natively embed the graph in its hardware.

When the hardware either does not have enough qubits, or requires long chains to embed an annealing problem, distributed quantum annealing can be a viable option. After finding an adequate graph split, the distribution fidelity $p_N(\delta)$ gives a lower bound for the probability of measuring a state that was evolved by annealing. Then, the average number of runs until the anneal state is measured is $[p_N(\delta)]^{-1}$. This means that the average time complexity (time that it takes to measure a state evolved by annealing) of the noisy protocol is $t_F [p_N(\delta)]^{-1}$. In addition, the average quantum communication complexity (how much of the entanglement resource is required by the protocol) is $M_D [p_N(\delta)]^{-1}$.

CONCLUSIONS

To summarize, we have introduced a protocol for distributed quantum annealing based on a combination of trotter decomposition and unitary distribution. We motivated the need for such a protocol by the increasing connectivity and size requirements that Quantum Annealing has to face in order to solve bigger and more complex problems. We showed that, using our protocol it is indeed possible to connect multiple annealers to
represents a bigger problem that wouldn’t fit on a single machine and successfully recover the global ground state with high probability. We have also investigated the robustness of our protocol in the presence of non-ideal distribution of entanglement and proved that it could, indeed, work under realistic assumptions. We find that the number of non-local operations on each Trotter step and the number of Trotter steps $M$ scale differently depending on how the graph is split, in the future it would be interesting to investigate if a recipe for optimal splitting can be achieved. The technology required to perform such distributions either in a local area network or at a larger scale are still in a proof of principle stage, but showing promise for the future. We thus conclude that distributed quantum annealing can be a viable way to augment the capabilities of a quantum annealer, when it cannot natively embed an annealing problem due to hardware limitations, either by its connectivity or by the number of required qubits. Having a cluster of quantum annealers connected with quantum communication channels could thus improve significantly the capability of solving larger and more complex optimization problems, as well as the possibility to explore optimization in a secure multi-party setting.

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Appendix A: Graph split: $n$-plication mechanism

The idea behind qubit $n$-plication in quantum annealing is to restrict the evolution of the $n$-plicated qubits in a way that they are kept aligned when measured. For instance, duplicating a qubit $q$ means to transform the label $q \rightarrow q_1, q_2$ and the state

$$|\psi\rangle_q = \alpha |0\rangle_q + \beta |1\rangle_q \rightarrow |\psi\rangle_{q_1, q_2} = \alpha |00\rangle_{q_1, q_2} + \beta |11\rangle_{q_1, q_2}$$

(A1)

This means that the annealing Hamiltonian $H(t)$ also has to be transformed. The transformed final Hamiltonian $H_F$ has to be such that when the labels $q_i$ are replaced by $q$, the original final Hamiltonian is obtained. For instance, the local field terms can be transformed as such

$$\sigma_q^z \rightarrow \frac{1}{n} \sum_{i=1}^{n} \sigma_{q_i}^z$$

(A2)

For the starting Hamiltonian $H_0$ the transformation is different. The ground state of the starting Hamiltonian corresponds to the superposition of all possible states that can be a solution of the optimization posed by $H_F$. For the $n$-plicated qubits we want to restrict the solution to $\{ |0\rangle^{\otimes n}, |1\rangle^{\otimes n} \}$, which means the initial state has to be

$$|GHZ_n\rangle = \frac{1}{\sqrt{2}} ( |0\rangle^{\otimes n} - |1\rangle^{\otimes n} )$$

(A3)

Then, for a system with the $n$-plicated qubit $q$, the starting Hamiltonian has to have the term

$$H_0 \sim \sigma_{q_1}^x \sigma_{q_2}^x \cdots \sigma_{q_n}^x = \prod_{i=1}^{n} \sigma_{q_i}^x$$

(A4)

This term has $|GHZ_n\rangle$ as one of the ground states. Because of this and because the action of $\sigma_0^x \sigma_1^z$ terms only change a relative phase in the superposition, the evolution by annealing will keep the $n$-plicated qubit in the space $\{ |0\rangle^{\otimes n}, |1\rangle^{\otimes n} \}$.

Appendix B: Trotterization convergence details

Since the local and non-local Hamiltonians do not commute, i.e. $[H_L(t), H_N(t)] \neq 0$ for $t \in [0, t_F]$, the Trotterization process will not be ideal. In the end, we want the Trotterized evolution to approximate quantum annealing. At step $k$ the evolution is given by

$$U_N^{(k)} U_L^{(k)} = \exp \{ -i H_N(t_k) \Delta t \} \exp \{ -i H_L(t_k) \Delta t \}$$

(B1)

The equivalent Hamiltonian $H_{eq}(t_k)$, when it exists, corresponds to the Hamiltonian that governs the evolution as if it was a continuous process governed by the Schrödinger equation. We want this equivalent Hamiltonian to converge to the annealing Hamiltonian $H(t)$ with increasingly smaller step sizes. The Baker-Campbell-Hausdorff expansion allows us to write an expression for $H_{eq}(t_k)$ at step $k$ [39]:

$$-i H_{eq}(t_k) \Delta t =$$

$$= \log \left[ \exp \{ -i H_N(t_k) \Delta t \} \cdot \exp \{ -i H_L(t_k) \Delta t \} \right]$$

$$= -i H(t_k) \Delta t + \sum_{n \geq 2} \sum_{|\omega|=n} g_\omega |\omega|$$

(B2)

where $\omega = \omega_1 \omega_2 \cdots \omega_{n-1}$ is a word of length $n = |\omega|$ and $\omega_i$ can be $X$ or $Y$ where

$$X = -i H_L(t_k) \Delta t,$$

$$Y = -i H_N(t_k) \Delta t.$$
\[ | \omega \rangle = [| \omega_0, \omega_1 \rangle, ... | \omega_{n-1} \rangle \]
represents nested commutators of the word \( \omega \) and \( g_\omega \) \( \text{is a scalar coefficient dependent on } \omega \).
In essence, we want that \( H_{eq}(t) \) to approximate the annealing Hamiltonian \( H(t) \). It then follows that we want to study the convergence properties of \( | |H_{eq}(t) - H(t)| | \).
With this aim, we use the relation eq. (B4) obtained by [35].

\[ ||X,Y||_L = ||X||_L ||Y||_L \]  \hspace{1cm} (B3)

\[ ||1/n \sum g_\omega | \omega \rangle ||_L \leq 2M^n/n \]  \hspace{1cm} (B4)

where
\[ M = \max (||X||_L, ||Y||_L) \]
\[ = \Delta t \cdot \max (||H_L(t_k)||_L, ||H_N(t_k)||_L) \]
\[ = \frac{\Delta t}{\Delta t_M} \]  \hspace{1cm} (B5)

Consequently, we obtain a bound of the Hamiltonian difference at step \( k \)
\[ ||H_{eq}(t_k) - H(t_k)||_L \Delta t \leq \sum_{n \geq 2} \frac{2}{n} \left( \frac{\Delta t}{\Delta t_M} \right)^n \]  \hspace{1cm} (B6)

where \( H(t) = H_L(t) + H_N(t) \) is the adiabatic Hamiltonian. Two observations can be made about this bound.
The first is that for \( \Delta t > \Delta t_M \), the absolute series diverges, meaning that the results of the Trotterization process cannot be predicted in this analysis. The series may conditionally converge for \( \Delta t > \Delta t_M \). In the case that \( H_{eq}(t) \) diverges, then the evolution cannot be described by the Schrödinger equation, and thus is not a continuous process. The second observation is that when \( \Delta t/\Delta t_M \ll 1 \), we have the bound
\[ ||H_{eq}(t_k) - H(t_k)||_L \leq \frac{\Delta t}{(\Delta t_M)^2} \]  \hspace{1cm} (B7)

Appendix C: Noisy distribution details

A generic distributed operation consists of a sequence of local unitaries and CNOT telegates. The noise is considered to come solely from the shared quantum state
\[ \rho_\Phi = x | \Phi \rangle \langle \Phi | + \frac{1-x}{4} I_\Phi , \]
where \( | \Phi \rangle = (|00\rangle + |11\rangle)/\sqrt{2} \) for the CNOT telegate. The CNOT distribution protocol shown in fig. 3 is equivalent to the circuit fig. 9, and so the CNOT telegate can be written as
\[ \text{DCNOT}[\rho] = \text{Tr}_\Phi \left[ U_{\text{DCNOT}}(\rho \otimes \rho_\Phi) U_{\text{DCNOT}}^\dagger \right] \]  \hspace{1cm} (C1)

with
\[ \Theta[\rho] = \frac{1}{4} \text{Tr}_\Phi \left[ U_{\text{DCNOT}}(\rho \otimes I_\Phi) U_{\text{DCNOT}}^\dagger \right] . \]

The CNOT[\rho] operation corresponds to the state after ideal distribution, without noise, while \( \Theta[\rho] \) is the state after a distribution fault occurs, i.e. \( \rho_\Phi = I_\Phi/4 \).

Consider \( D_k[\rho] \) the state that results from the distribution of the initial state \( \rho \), with all possible \( k \) distribution faults in \( M_D \) distribution steps. We have that \( Tr D_0[\rho] = 1 \). For example, a state \( \rho_2 \) which has to use 2 noisy DCNOT operations can be rewritten as
\[ \rho_2 = x^2 D_0[\rho_0] + 2x(1-x) D_1[\rho_0] + (1-x)^2 D_2[\rho_0] . \]  \hspace{1cm} (C2)

\[ D_0[\rho_0] = U_2 \circ CNOT \circ U_1 \circ CNOT \circ U_0[\rho_0] \]
\[ D_1[\rho_0] = \frac{1}{2} U_2 \circ \Theta \circ U_1 \circ CNOT \circ U_0[\rho_0] \]
\[ + \frac{1}{2} U_2 \circ CNOT \circ U_1 \circ \Theta \circ U_0[\rho_0] \]
\[ D_0[\rho_0] = U_2 \circ \Theta \circ U_1 \circ \Theta \circ U_0[\rho_0] \]

with \( U_i[\rho] = U_i[\rho] U_i^\dagger \) being an arbitrary unitary gate operation (noiseless) and \( f \circ g[x] = f[g[x]] \) being the composition operation. Given that the probability of a perfect distributed operation (telegate) is given by \( x \), then the state \( \rho_0 \) after \( M_D \) distribution steps is given by the binomial distribution:
\[ \rho_{MD} = \sum_{k=0}^{M_D} \left( \frac{M_D}{k} \right) x^{M_D-k} (1-x)^k D_k[\rho_0] \]  \hspace{1cm} (C3)

Writing \( F_\Phi = 1 - \delta \), we have \( x = 1 - \frac{4}{3} \delta \). After \( M_D \) DCNOT operations, the fidelity between the states after noisy (\( \rho_{MD} \)) and noiseless (\( D_0[\rho_0] \)) distribution us \( p_N(\delta) \), and is lower bounded by the first term in the sum
\[ p_N(\delta) \geq x^{M_D} = \left( 1 - \frac{4}{3} \delta \right)^{M_D} , \]  \hspace{1cm} (C4)
With $|\Psi\rangle = |D_0\rangle |\rho_0\rangle$ being the state after ideal (noiseless) distribution, $p_N(\delta)$ is given by

$$p_N(\delta) = \langle \Psi | \rho_{MD} | \Psi \rangle = \sum_{k=0}^{MD} \binom{MD}{k} x^{MD-k} (1-x)^k f_k,$$

with $f_k = \langle \Psi | D_k | \rho_0 \rangle | \Psi \rangle$ being a coefficient which depends on the number of faults in the distribution $k$. Assuming that successive faults reduce the probability of measuring $|\Psi \rangle$ by $\beta$, we get the ansatz

$$f_k = \beta^k, \quad \beta < 1$$

where $\beta = \langle \Psi | D_1 | \rho_0 \rangle | \Psi \rangle$. With this ansatz, $p_N(\delta)$ can be reduced to a closed-form expression using the binomial theorem, and thereby obtaining eq. (10), reproduced here.

$$p_N(\delta; \beta) = \left(1 - \frac{4}{3} \delta (1-\beta)\right)^{MD}$$


