

Hamiltonian Simulation Circuit

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Abstract

In this work we deal with the problem of designing efficient circuit to simulate the development in time of the initial state by the given Δ -local Hamiltonian. We suggest an heuristic and compare the circuit depth obtained by them to the circuit which is yielded by naive approach.

The problem. Generate a circuit, using no more than 10 qubits, that approximates the unitary e^{-iH} where H is the qubit Hamiltonian of a **LiH** (lithium hydride) molecule. The **LiH** Hamiltonian is composed of 276 Pauli strings, and can be found [HERE](#). The approximation error is defined in the next section, and should be less than 0.1. The circuit should be composed of the CX and single qubit gates only.

Definition. Let H_i be a single term of the Hamiltonian, we will define the $\text{support}(H_i)$ to be a vector $v \in \mathbb{F}_2^n$ such that $v_j = 1$ if H_i act non-trivially on the j th qubit and 0 else.

Notice that each term H_i such that $\text{support}(H_i) = 1^a 0^b = (1, 1, \dots, 1, 0, \dots, 0)$ can be simulated in parallel with term H_j with complementary support (i.e $0^a 1^b$). In addition, if the support isn't a continuous segment of 1's then one could perform rotate the base by permutation at a cost of the "mixing size". [\[COMMENT\]](#) Change G to be the graph which each node $v = (P, \text{base change})$.

Definition. Denote by m the number of local Hamiltonian terms and consider the graph $G = (\{H_i\}, E)$ which is a weighted complete graph over m vertices. Such that the weight function defined by $w(u, v) = \# \{j | H_v^j \neq H_u^j \text{ and } H_v^j, H_u^j \neq I\}$. We will call to such G the Qubic embedding or shortly just G .

Theorem. Given a Δ -local Hamiltonian over n qubits and m terms such that $n > \frac{m}{\Delta+1}$ then there is a circuit which simulate a single Trotterize step at depth:

$$\begin{aligned} D(n, m) &\leq \log_{\frac{\Delta+1}{\Delta}}(m) \cdot \frac{\Delta}{1+\Delta} n \cdot (2 + 2\Delta) \\ &= \log_{\frac{\Delta+1}{\Delta}}(m) \cdot 2\Delta n \end{aligned}$$

Let us abuse the notation and denote by H also the random variable which count the support size of local term which is sampled uniformly from the total Hamiltonian.

Theorem. [\[COMMENT\]](#) Change to uniform drawn Δ -local Hamiltonian. Given an Hamiltonian with expected support $\mathbb{E}[H] := \mathbb{E}_{\sim_i}[\text{Ham}(\text{support}(H_i))]$ then there is a circuit which simulate a single step $e^{iH\Delta t}$ at depth:

$$\begin{aligned} D(n, m) &\leq m \left(1 + \mathbb{E}_{\sim_i}[H] \right) \\ &\quad + 4w(\text{MST}(G)) + \left(2 - \frac{1}{m} \right) n^2 \end{aligned}$$

Where $w(\text{MST}(G))$ is the weight of the minimal spanning tree of G . Please, notice that in our case $n = 10, m = 276$.

Before continuing to the proof let's analyze the naive approach, just handle each of Hamiltonian terms one by one. Assume the the hamming weight of the support of H_i equal to d_i . so we pay two steps to rotate each wire into the parity base and uncompute it in the end. then we could apply the CX from each qubit in the support to a chosen one which will sum the wires parity and finally will propagate the sum through the $RZ(\theta\Delta t)$ gate (rotation by the coefficient of the term and the step size). So in total we pay $2 + 2d_i$ for each term and therefore the whole circuit will require:

$$D^{\text{naive}}(n, m) = \sum_i 2 + 2d_i = m \sum_i \frac{2 + 2d_i}{m} = 2m \left(1 + \mathbb{E}_{\sim_i}[H] \right)$$

Hence the theorem inequality could be written as:

$$D(n, m) \leq \frac{1}{2} D^{\text{naive}}(n, m) + 4w(\text{MST}(G)) + \left(2 - \frac{1}{m} \right) n^2$$

And therefore, for an Hamiltonians which have a grater number of terms than the number of the qubits which they act on, we expect to see an improvement which is almost an half of the naive solution. In fact this improvement is much more significant than might seem as simulating up to ε -error requires **Poly** $\left(\frac{1}{\varepsilon}\right)$ repetitions[[COMMENT](#)] [add a citation](#), so for full simulation we have obtained a $\sim \left(\frac{1}{2}\right)^{\text{poly}\left(\frac{1}{\varepsilon}\right)}$ speed-up factor.

Solution. Let's formulate the claim which mentioned earlier, suppose that H_i, H_j are terms which their supports are disjoint and also they lay on continues segments. Then we can simulate the terms in parallel.

Lemma. *let H_i, H_j and integers $a, b, c, da < c, d < b$ such the supports of the terms are $0^a 1^b, 1^c 0^d$. Then we can simulate $e^{iH_i + H_j}$ with circuit with depth $2 + \max_{i,j} 2 \cdot \mathbf{Ham}(\text{support}(H_l))$.*

Proof. It's clears that the first rotation, and the final uncompute stages remain unchanged. So we should deal only with 2-qubits operators (the cnot's). By the $a < c$ restriction it follows that each CX gate in the naive implementation of the H_i cross only elements which H_j act trivially on. Hence if at time-step k the original operations in each of the naive implementations were $CX(i_1, i_2), CX(j_1, j_2)$ then $j_1, j_2 \notin [i_1, i_2]$. Repeating on the above argument, but consider the $d < b$ constraint instead yields that $i_1, i_2 \notin [j_1, j_2] \Rightarrow [i_1, i_2] \cap [j_1, j_2] = \emptyset$ and therefore we can put those gates on the same time step \square

Corollary. Assume that H_i, H_j have a disjoint supports. Then by swapping the wires (including phase swap) into continues segments, we obtain a simulation at cost of : $\text{cost}(\text{Permutation}) + 2 + \max_{i,j} 2 \cdot \mathbf{Ham}(\text{support}(H_l))$.

The permutation which swapping between $v_1, v_2 \dots v_k \leftrightarrow u_1, u_2 \dots u_k$ where $v_i, u_j \in [n]$ are wires/qubits indices could be done by swapping each of the pair separately. Namely, **SWAP** (v_i, u_i) for each of the pairs. Figure 1 presents the two qubit gate **SWAP** and Figure 2 is an example of the generalization obtained by concatenation such gates.

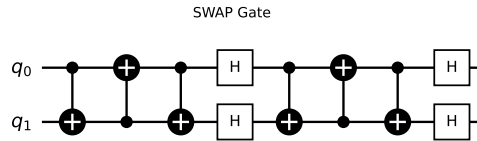


Figure 1: The implementation of the two-qubit **SWAP** gate. The left part replace the bit values of the qubits, then we are rotate into the dual space and swap the phases finally we return back to the original space.

Corollary. Let $P = \langle v_1, v_2 \dots v_l \rangle$ a vertices path in the graph G as defined above. And let U be circuit such that for each $v_i \in P$ U apply the circuit U_i at the [\[COMMENT\] I should define it](#) base. Then there exist an implementation for that circuit at depth:

$$\begin{aligned}
D(U) &= \sum_i D(U_i) + D(\mathbf{SWAP}^{v_i, v_{i+1}}) \\
&= \sum_i D(U_i) + \mathbf{Ham}(v_{i-1}, v_i) \cdot \text{Depth}(\mathbf{SWAP}) \\
&= \sum_i D(U_i) + w(v_{i-1}, v_i) \cdot \text{Depth}(\mathbf{SWAP}) \\
&= w(P) \cdot \text{Depth}(\mathbf{SWAP}) + \sum_i D(U_i)
\end{aligned}$$

Expectation of intersect terms. Denote by X_i^j the indicator which assign 1 to the event that the i term act non trivially on the j th qubit. Then it's clear that

$$\mathbb{E}[\text{intersects}] \leq \mathbb{E} \left[\sum_{i,k}^m n \cdot X_i^j X_k^j \right] = \binom{m}{2} np^2$$

If we sample the terms from a uniform distribution over the Δ -local terms then $p = \binom{n-1}{\Delta-1} / \binom{n}{\Delta} = \frac{\Delta}{n}$. And therefore the expectation of the intersections is less than $\binom{m}{2} \frac{\Delta^2}{n}$. Hence by markov inequality we obtain that

$$\Pr \left[\text{intersects} \geq \frac{1}{2} \binom{m}{2} \right] \leq \frac{2\Delta^2}{n}$$

The Algorithm.

The general concept. We will identify each pair of

Algorithm 1: Simulate a single step $e^{iH\Delta t}$

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1
2  $|\psi\rangle \leftarrow$  the initial state.  $T \leftarrow \text{MST}(G)$ .
3  $P \leftarrow$  ordering of the vertices  $\langle v_1, \dots, v_l \rangle$  according to the DFS scanning of  $T$ .
4
5 for  $v \in P$  do
6    $|\psi\rangle \leftarrow \mathbf{SWAP}^{v_{i-1}, v_i}$ 
7   while  $\exists H_i, H_j$  s.t  $\text{supp}(H_i) = v = \overline{\text{supp}(H_j)}$  do
8      $|\psi\rangle \leftarrow U(H_i)_v \otimes U(H_j)_{\bar{v}} |\psi\rangle$ 
9     mark  $H_i, H_j$ .
10  end
11 end
12
13 complete the left terms by the naive approach.
14 return  $|\psi\rangle$ 

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Lemma. Let $\{X_i\}$ be a monotonic set of numbers bounded by 2μ with expectation μ and variance σ^2 . Then:

$$\sum \frac{1}{m} \max \{X_i, X_{i+\frac{n}{2}}\} \leq \mu + \frac{1}{2}\sigma^2$$

Proof. Define $Y : \Omega \rightarrow \mathbb{R}$ to be the random variable which is the hist of the X by μ i.e $Y = X - \mu$. From the linearity of expectation we get that $\mathbb{E}[X] = \mu + \mathbb{E}[Y]$. let's note by Y^s and Y^{as} the symmetric and asymmetric parts of Y . the \square .

$$\mathbb{E}[X] = \frac{1}{2}M + \mathbb{E}[X^s - M] = \frac{1}{2}M + \mathbb{E}[(X - M)^s + (X - M)^{as}] = \frac{1}{2}M + \frac{1}{2}\mu + \mathbb{E}[(X - M)^{as}]$$

Lemma. Let $H = \frac{1}{m} \sum H_i$ be Hamiltonian such each term H_i is a generalized pauli operator. Let P be random pauli which sampled uniformly. Then:

Lemma. Majority principle. We can always assume that the more then quarter of the paulis are identity. [\[COMMENT\]](#) doesn't correct, but can be passed if we change the definition of the solid tensor to include only differ coordinates.

$$U^t = \sqrt{X}^\dagger \sqrt{X} U \sqrt{X} \sqrt{X}^\dagger U \sqrt{X}^\dagger \sqrt{X} \dots U \sqrt{X}^\dagger \sqrt{X}$$

1 [\[COMMENT\]](#) Enter it

Lemma I A graph with all degrees at most d satisfies:

$$\alpha(G) \geq \frac{|V(G)|}{d+1}$$

Proof. Denote by $U \subset V$ the maximal independent subset of V . By the maximality it follows that for every $w \in V/U$ there is at least one neighbor in U (otherwise the $U \cup w$ is an independent subset with a bigger size). Therefore

$$|V/U| = |\Gamma(U)/U| \leq |\Gamma(U)| \leq d|U| = d\alpha(G)$$

Combining the fact that $|V| = |U| + |V/U|$ we get that:

$$\begin{aligned} \Rightarrow |V| &= |U| + |V/U| \leq \alpha(G) + d\alpha(G) \\ \Rightarrow \alpha(G) &\geq \frac{|V(G)|}{d+1} \end{aligned}$$

□

Lemma. Bravyi Kitaev Transformation Degree. [\[COMMENT\]](#) Prove that, in the $SP(G)$ Graph the degree is bounded by $4^{4 \log(n)} - 2^{4 \log n}$

$$\begin{aligned} m_{i+1} &\leftarrow m_i - \alpha(G) \\ m_{i+1} &\leftarrow \left(1 - \frac{1}{1+\Delta}\right) m_i \\ \left(1 - \frac{1}{1+\Delta}\right)^h m_0 &= 1 \\ h &= \log_{\left(\frac{1+\Delta}{\Delta}\right)}(m_0) \end{aligned}$$

Lemma. For any $q \in \mathbb{N}$

$$\frac{1}{2} (1 + q + \dots q^{n-1}) + q^n \leq \left(1 - \frac{1}{2(q+1)}\right) \frac{q^{n+1} - 1}{q - 1}$$

$$\begin{aligned} f(m) &= f\left(\frac{1}{2}\alpha(m)\right) + f(m - \alpha(m)) \\ &\leq f\left(\frac{1}{2(\Delta+1)}m\right) + f\left(\left(1 - \frac{1}{(\Delta+1)}\right)m\right) \end{aligned}$$

$$\begin{aligned} f(m) &= \xi + f(m - \alpha(m)) \\ &= 2\xi + f(m - \alpha(m) - \alpha(m - \alpha(m))) \\ &\leq 2\xi + f(m - \alpha(m) - \alpha(m - \alpha(m))) \end{aligned}$$

Randomization Trick.