Quantum Circuit Transformation Based on Subgraph Isomorphism and Tabu Search*

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Abstract. The goal of quantum circuit adjustment is to construct mappings from logical quantum circuits to physical ones in an acceptable amount of time, and in the meantime to introduce as few auxiliary gates as possible. We present an effective approach to constructing the mappings. It consists of two keys steps: one makes use of a combined subgraph isomorphism and complement (CSIC) to initialize a mapping, the other dynamically adjusts the mapping by using a Tabu search based adjustment (TSA). Our experiments show that, compared with the wghtgraph recently considered in the literature, CSIC can save 22.26% of auxiliary gates and reduce the depths of output circuits by 11.76% on average in the initialization of the mapping, and TSA has a better scalability than many state-of-the-art algorithms for adjusting mappings.

Keywords: Quantum circuit transformation \cdot Subgraph isomorphism \cdot Initial mapping \cdot Tabu search

1 Introduction

The entanglement of a quantum system with its surrounding environment will lead to quantum decoherence. It is unrealistic to use quantum error correction in the circuit mapping process, since there are only dozens of available qubits for quantum devices in the NISQ era [16]. It is necessary to transform circuits by adding auxiliary gates to satisfy both logical and physical constraints, since quantum algorithms do not consider any hardware connectivity constraints. Hence, quantum circuit transformation is an important part of quantum circuit compilation. We need a set of highly efficient and automatic mapping procedures and adjustment routines to perform circuit transformation. In this process noise may be introduced, which brings a huge challenge to circuit compilation because noise has a significant impact on the final circuit and may make the result meaningless.

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In the current work, we adjust the lifetime of qubits through parallelization, and use SubgraphMatching [20] to generate partial isomorphic subgraphs of logical circuits LC and physical circuits PC as part of the initial mapping. The advantage of the initial mapping result is that we use the appropriate subgraph isomorphism and the two-way connection of the logical circuit and the physical circuit to obtain a dense initial mapping, which avoids certain nodes from being mapped to remote locations. We use Tabu search [6] to generate circuits that can be executed on physical devices. Tabu search can avoid falling into local optimum and swapping the recently swapped qubits, thereby improve the parallelism of quantum gates. We add the SWAPs associated with the gates on the shortest path to the candidate set, which greatly reduces the search space and improves the search speed. Our heuristic function not only considers the current gates but also the constraints of the gates already considered.

We compared CSIC with the start-of-art initial mapping method wghtgraph [10] and optm [25]. On average, the auxiliary gates of the CSIC algorithm are reduced by 22.44%, 27.02%, and the depths are reduced by 11.25%, 14.12%, respectively. We Compared with wgtgraph [10] and SABRE [9], the adjustment algorithm is compared. In contrast, TSA can deal 159 circuits in a few minutes, but other algorithms are difficult to handle middle or large circuits. The number of SWAP gates added by wgtgraph is 26.87% (resp. 24.89%) smaller than TSA_{num} (resp. TSA_{dep}) on average. SABRE tested 159 circuits and only 29 circuits were successfully mapped within the five-minute limit.

The main contributions of this paper are as follows.

- 1. We propose to use the combined subgraph isomorphism algorithm to generate part of the initial mapping and then complete the mapping based on the connectivity between qubits.
- 2. We propose a heuristic circuit adjustment algorithm based on Tabu search (TSA) [6], which can handle large circuits in a shorter time at a lower cost, compared with existing precise search and heuristic algorithms.
- 3. We propose a look-ahead heuristic function that considers both the current gates and the gates yet to be processed. It filters out a swap that is beneficial to the current gates and also brings closer the gates to be processed.
- 4. We tested 159 circuits, and the results show that the initial mapping generated by our method can make the subsequent insertion of fewer gates, and the adjustment algorithm can be extended to medium and large circuits.

The rest of this paper is organized as follows. In Section 2 we introduce some related works. In Section 3 we recall some background of quantum computing and quantum information. In Section 4 we introduce the problem of the transformation of quantum circuits. In Section 5 we describe and analyze our algorithm in detail. The experimental results are reported in Section 6. The last section concludes the paper and discusses some future research.

2 Related work

Quantum technology has been applied in practice, but large quantum computers have not yet been built. Most of the contributions of quantum information to computer science are still in the theoretical stage. In 2017, IBM developed the first 5-qubit backend called IBM QX2, followed by the 16-qubit backend IBM QX3. The revised versions of them are called IBM QX4 and IBM QX5, respectively. IBM Q Experience provides the public with free quantum computer resources on the cloud and opens source the quantum computing software framework Qiskit⁴.

Users of these early quantum computers mainly rely on quantum circuits to implement quantum algorithms. They design logical circuits which then go through the step of circuit transformation in order to map logical qubits to physical ones before the logical circuits are exectued in physical devices. A big challenge for quantum information is the problem of quantum decoherence. Due to the decoherence of qubits, quantum gates need to be applied in a coherent period as the time for a qubit to stay in a coherent state is very short. The longest coherence time of a superconducting quantum chip is still within 10us-100us.

There are several initial mapping methods. Paler [15] showed that the initial mapping has an important influence on quantum circuit transformation. He proposed a heuristic method to find the initial mapping. Just by placing qubits in different positions from the default in the actual circuit instance on the actual NISQ device, the time cost can be reduced by up to 10%. Li et al. [9] proposed a novel reverse traversal technique, which determines the initial mapping by considering the entire circuit. Zhou et al. [23] put forward an annealing algorithm to find an initial mapping, but it is unstable. In [10], Li et al. considered the subgraph isomorphism algorithm wghtgraph to generate an initial mapping, which is currently the most start-of-art, so we will compare it.

The goal of circuit adjustment algorithm is to find a minimum number of SWAPs. There are currently five main methods for solving the quantum circuit adjustment problem.

Unitary matrix decomposition algorithm. It is used in [8, 14] to rearrange the quantum circuit from the beginning while retaining the input circuit. It can be applied to a broader class of circuits consisting of generic gate sets, but the results are not as efficient as a compiler designed specifically for this task.

Converting into some existing problems. This approach converts the quantum circuit transformation problem into some existing problems, such as AI planning [22,3], Integer Linear Programming (ILP) [1], or Satisfiability Modulo Theories (SMT) [12]. Existing tools for those problems are then used to find acceptable results. The approach cannot take advantage of certain properties of quantum mapping, which is a drawback. Furthermore, as the time cost is usually long, it can only deal with small quantum circuits.

Exact methods. Siraichi et al. [19] have proposed an exact method. It will iterate all possible mappings for all dependencies, so it is only suitable for simple

⁴ https://www.qiskit.org/

quantum architecture and cannot be extended to complex quantum architectures.

Graph theory. In [17], Shafaei et al. have used the minimum linear permutation solution in graph theory to model the problem of reducing the interaction distance. The idea is to first divide a given circuit into several subcircuits and apply the minimum linear permutation solution, respectively. Then we turn non-adjacent gates in the subcircuits into adjacent gates by adding auxiliary gates. Finally, we can use the minimum linear permutation solution to find an appropriate permutation and use bubble sort to calculate the number of necessary SWAP gates. In [7,11], a two-step method is used to reduce the quantum circuit transformation to the graph problem to minimize the number of auxiliary gates, based on the graph coloring problem and the largest subgraph isomorphism problem.

Heuristic search. Heuristic search uses an evaluation function to obtain an acceptable solution in exponential time. Zulehner [25] suggests to layer the circuits, then determines compatible mappings for each of these layers to add as few auxiliary gates as possible. Zhou et al. [23] have designed a heuristic search algorithm with a novel selection mechanism. Instead of choosing the operation with the lowest cost to apply, one can look forward one step and then choose the best continuous operation. In this way, the algorithm can effectively avoid local minimum. Moreover, a pruning mechanism is introduced to reduce the search space's size and ensure that the program terminates in a reasonable amount of time.

Li et al. [9] have proposed a SWAP-based search algorithm SABRE. Compared with previous search algorithms based on exhaustive mapping, SABRE ensures the scalability of SABRE to adapt to large quantum circuits in the NISQ era. In [4], a routing algorithm called $t\,|ket\rangle$ ensures that any quantum circuit can be compiled into any architecture. The algorithm is divided into four stages: decomposing the input circuit into time steps, determining the initial mapping, routing across time steps, and finally cleaning up. The heuristics in $t\,|ket\rangle$ give the same or better results than other circuit transformation systems in terms of the depth and the total number of gates in the compiled circuit, with much shorter running time, and can handle larger circuits. In [21], a variation-aware qubit movement strategy is proposed. It takes advantage of the change in error rate and a change-aware quantum circuit transformation strategy by trying to select the route with the lowest probability of failure. This strategy uses the error rate of SWAPs to allocate logical qubits to physical qubits, thus avoiding paths with high error rates as much as possible.

3 Background

This section introduces some notions and notations of quantum computing and quantum information.

Qubits Classical information is stored in bits, while quantum information is stored in gubits. Besides two basic states $|0\rangle$ and $|1\rangle$, a qubit can be in any

linear superposition state like $|\phi\rangle=a\,|0\rangle+b\,|1\rangle$, where $a,b\in\mathbb{C}$ satisfy the condition $|a|^2+|b|^2=1$. The intuition is that $|\phi\rangle$ is in the state $|0\rangle$ with the probability $|a|^2$ or in the state $|1\rangle$ with the probability $|b|^2$. We use letters \mathbf{q} and q in different fonts to represent physical qubits and logical qubits.

Quantum Gates By inserting quantum gates, the state of the qubits is changed. Some commonly used quantum gate symbols and their semantics in terms of matrices are shown in Fig. 1.

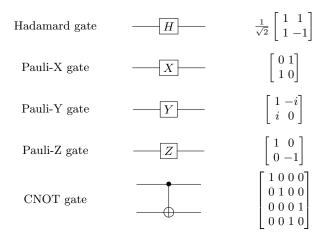


Fig. 1. The symbols of common quantum gates and their matrices

Quantum Circuit A quantum logical circuit consists of quantum gates interconnected by quantum wires [5]; see Fig. 2 for an example. A quantum wire is a mechanism for moving quantum data from one location to another. Each line represents a qubit, and the gates on the lines act on the corresponding qubits. The execution order of a quantum logical circuit is from left to right. The width w of a circuit refers to the number of qubits in the circuit. The depth d of a circuit refers to the number of layers executing in parallel. For example, the depth of the circuit in Fig. 2 is 6, and the width is 5. In this paper, a circuit with a depth less than 100 is a called a small circuit, a circuit with a depth greater than 1000 is called a large circuit, and the rest are medium circuits. It is unnecessary to consider quantum gates acting on single qubits in circuit adjust, since the single qubits are local [18].

Architectures We mainly discuss the physical circuits of IBM Q series. Let $\mathcal{AG}_{\mathcal{P}} = (V_P, E_P)$ denote the architecture graph of the physical circuit, where V_P denotes the set of physical qubits and E_P denotes the set of edges that connect CNOT gates. In Fig. 6, (a) and (b) are the physical architecture graphs

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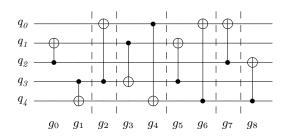


Fig. 2. Original circuit

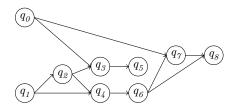


Fig. 3. The directed acyclic graph (DAG) of original circuit in Fig. 2.

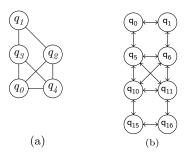


Fig. 4. (a) The architecture graph of original circuit in Fig. 2. (b) The partial architecture graph of IBM Q20.

of the 5-qubit IBM QX2 and IBM QX4, respectively. (d) and (c) are the physical architecture graphs of the 16-qubit IBM QX3 and IBM QX5, respectively, and (e) is the physical architecture graph of IBM Q20. The direction in each edge indicates the control direction of a 2-qubit gate, and 2-qubit gates can only be performed between qubits with edges connected. IBM physical circuits only support single quantum gates and CNOT gates between two adjacent qubits.

Given a logical circuit LC, a physical structure \mathcal{AG}_P , an initial mapping τ , and a CNOT gate $g = \langle q_i, q_j \rangle$, where q_i is the control qubit, q_j is the target qubit. If gate g is executable on a physical circuit, then $\langle \tau(q_i), \tau(q_j) \rangle$ must be a directed edge on \mathcal{AG}_P .

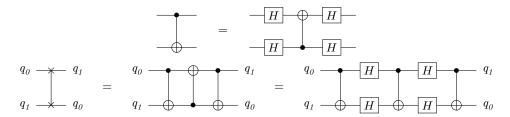


Fig. 5. The above circuits changes the direction of the CNOT gate by adding four H gates, and three equivalent circuits of SWAP gate are given below.

Example 1. Fig. 4 (a) is the logical structure of Fig. 2. Fig. 4 (b) is the partial architecture graph of IBM Q20. An initial mapping is

$$\tau = \{q_0 \rightarrow \mathsf{q}_{10}, \ q_1 \rightarrow \mathsf{q}_0, \ q_2 \rightarrow \mathsf{q}_6, \ q_3 \rightarrow \mathsf{q}_5, \ q_4 \rightarrow \mathsf{q}_{11}\}.$$

The 2-qubit gate $g_0 = \langle q_2, q_1 \rangle$ is not executable, since the edge $\langle \tau(q_2), \tau(q_1) \rangle = \langle \mathsf{q_6}, \mathsf{q_0} \rangle$ does not exist in \mathcal{AG}_P . But $g_3 = \langle q_1, q_3 \rangle$ is executable, since the edge $\langle \tau(q_1), \tau(q_3) \rangle = \langle \mathsf{q_0}, \mathsf{q_5} \rangle$ exists in \mathcal{AG}_P .

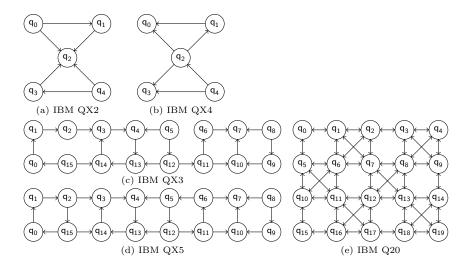


Fig. 6. IBM QX architectures

4 Problem Description and Solution Outline

Before the circuit transformation we are supposed to given a circuit that has only single quantum gates and CNOT gates [2, 13]. We add auxiliary gates (see Fig. 5)

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to move two non-adjacent quantum positions to adjacent positions or change the direction of a CNOT gate, but this process may introduce errors. We hope to find a circuit transformation algorithm that can produce the output circuit with a minimal number of auxiliary gates and a small circuit depth in an acceptable amount of time. A quantum circuit transformation mainly includes the following four steps. Isomorphism and adjustment are both NP-complete [19].

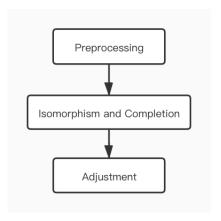


Fig. 7. Circuit transformation process

- 1. Preprocess the logical quantum circuit. This step includes extracting the logical architecture graph of the circuit, adjusting the life cycle of qubits as done in [24]), and calculating the shortest paths of the physical circuit.
- 2. Compute isomorphic substructures and generate a high-quality initial mapping. This step uses the subgraph isomorphism algorithm to find part of the initial mappingas done in [20]. Then we perform a mapping completion to include the remaining nodes that do not satisfy all isomorphism requirements, according to the connectivity between the unmapped node and the mapped nodes.
- 3. Transform logical circuits to meet physical constraints. Circuit transformation need to be performed before compiling quantum circuits, since the design of quantum algorithms usually does not refer to the connectivity constraints of any hardware. Therefore, it is indispensible for any quantum compiler.

5 Solution Details

Our solution includes preprocessing, isomorphism, completion, and circuit adjustment, which will be presented in turn.

5.1 Preprocessing

Before a circuit adjustment, we can preprocess it by collect my data so as to shorten our search time and space. In the preprocessing step, we adjust the input circuit described by an openQASM program to shorten the life cycle of qubits. Then we use Breadth-First Search (BFS) to calculate the shortest distance between each node on the architecture graph.

We use a layered method to analyze the life cycle of qubits and pack the gates that can be executed in parallel into a bundle, forming a layered bundle format [24]. A conversion method is designed to use the layered bundle format to determine which gates can be moved, which reduces the life cycle of these qubits. The algorithm reduces the error rate of quantum programs by 11% on average. In most quantum workloads, the longest qubit lifetime and the average qubit lifetime can be reduced by more than 20%, and the execution time of some quantum programs can also be reduced.

Shortest Distance Given a physical architecture graph and assume the distance of each edge is 1, we can use Floyd-Warshall algorithm to calculate the shortest distance matrix dist[i][j], which represents the shortest distance from \mathbf{q}_i to \mathbf{q}_i .

For IBM QX2, QX3, QX4, QX5, a SWAP needs 7 gates (3 CNOT gates and 4 H gates). Only 4 H gates are needed to change the direction of an adjacent CNOT gate. For a CNOT gate $g = \langle q_i, q_j \rangle$, the two qubits are mapped to \mathfrak{q}_m and \mathfrak{q}_n , respectively, with $\tau(q_i) = \mathfrak{q}_m$, $\tau(q_j) = \mathfrak{q}_n$. Then the cost of executing g under the shortest distance path is $cost_{cnot}(q_i, q_j) = 7 \times (dist[m][n] - 1)$. For IBM Q20, in which all edges are bidirected, a SWAP requires 3 CNOT gates. Thus the cost between them is $cost_{cnot}(q_i, q_j) = 3 \times (dist[m][n] - 1)$. The time complexity is $O(N^3)$.

Example 2. Take the QX5 (see in Fig. 6 (d)) structure as an example. Suppose there is a CNOT gate $g = \langle q_1, q_2 \rangle$, q_1 is mapped to q_1 , q_2 is mapped to q_{14} , and the shortest distance between them is dist[1][14] = 3. There are 3 shortest paths to move q_1 to the adjacent position of q_{14} : $\Pi = \{\pi_0, \pi_1, \pi_2\}$, $\pi_0 = \mathsf{q}_1 \to \mathsf{q}_2 \to \mathsf{q}_3 \to \mathsf{q}_{14}$, $\pi_1 = \mathsf{q}_1 \to \mathsf{q}_2 \to \mathsf{q}_{15} \to \mathsf{q}_{14}$, $\pi_2 = \mathsf{q}_1 \to \mathsf{q}_0 \to \mathsf{q}_{15} \to \mathsf{q}_{14}$. Their costs are given by $cost_{\pi_0} = 18$, $cost_{\pi_1} = 14$, and $cost_{\pi_2} = 14$, respectively.

Circuit Layering Quantum gates acting on different qubits can be executed in parallel. Therefore, we classify the gates that can be executed in parallel into one layer, otherwise we add a new layer. The notation $L(LC) = \{\mathcal{L}_0, \mathcal{L}_1, ..., \mathcal{L}_n\}$ represents the layered circuit, where \mathcal{L}_i $(0 \le i \le n)$ stands for a quantum gate set that can be executed in parallel. The quantum gate set separated by the dotted line in Fig. 2 are the following $\mathcal{L}_0 = \{g_0, g_1\}, \mathcal{L}_1 = \{g_2\}, \mathcal{L}_2 = \{g_3, g_4\}, \mathcal{L}_3 = \{g_5, g_6\}, \mathcal{L}_4 = \{g_7\}, \mathcal{L}_5 = \{g_8\}.$

At the same time, we generate logical circuit architecture graph $\mathcal{AG}_{\mathcal{L}} = (V_L, E_L)$, which is an undirected graph with V_L being the set of vertices, and E_L representing the set of undirected edges that the CNOT gates can execute.

5.2 Isomorphism and Completion

We use the subgraph isomorphism algorithm to find a partial initial mapping that promissingly minimize auxiliary gates added by the output circuit.

As it is known, not all logical architecture graph have a subgraph that exactly matches all the nodes in physical architecture graph. In a physical architecture graph, it is almost impossible to find a subgraph that exactly matches all the nodes in a logical architecture graph. Thus, we regard the mapping with the largest number of matching nodes as the partial mapping. SubgraphMatching compares several state-of-the-art subgraph isomorphism algorithm compositions. It shows that the best performance can be achieved by using filters and the sorting ideas of the GraphQL algorithm to process candidate nodes, and the local candidates calculation method LFTJ based on set-intersection to enumerate the results. We artificially connect the isolating qubits to the qubits with the largest degree in the logical architecture graph, since SubgraphMatching cannot handle disconnected graphs. We would like to minimize the impact of logical dependency graphs, thus map isolated nodes to nodes with the largest degree.

The input of Algorithm 1 is a target graph (\mathcal{AG}_P) , a query graph (\mathcal{AG}_L) , and the partial mapping set T. First, we initialize an empty queue Q. Then we traverse τ and add the unmapped nodes to the queue Q. For the unmapped nodes, we try to map them to the vicinity of the mapped nodes in \mathcal{AG}_P . $q \to -1$ means that q is not mapped to any physical node, so we need to perform a mapping completion. Finally, we generate a dense mapping, which can reduce the added auxiliary gates. We could try to match the remaining unmapped nodes randomly, but it may lead to a mapping with a node far away from other nodes. If an unmapped node has an edge adjacent to a matched node in the query graph, it will be matched to one of the adjacent nodes first. In this way, we can obtain all initial candidate mappings.

In Algorithm 1, Line 2 selects the partial mapping with the most mapped nodes l as the candidate set. Lines 3-28 complete the partial mapping. In Line 6, we initialize an empty queue Q, which stores unmapped logical qubits. In Lines 8-10, we traverse the mapping τ and add the unmapped qubits to Q. We iterate the loop until Q is empty, and all logical qubits are mapped to physical qubits. We take out the first element in Q to q. Lines 13-14 are used to get the adjacency matrices of \mathcal{AG}_P and \mathcal{AG}_L , respectively. Line 15 initializes an empty map cans, sort by a descending order of whether q_m is mapped and the number of gate appearances. Line 16 store the nodes q_m connected to q in the adjacency matrix in cans. Lines 17-28 traverse cans, select the node q_m that has been mapped to the node (cans.first) on physical architecture graph, and has the largest number of connections to q in cans, with q being the node that q_m is mapped to in τ . Line 20 deletes the node from cans. Lines 21-26 select the node k adjacent to q in the adjacency matrix, and maps q to that node.

Example 3. Following the previous example, we first use the CSIC algorithm for the logical architecture graph given in Fig. 4 (a) and the physical architecture graph given in Fig. 6 (e) to obtain the partial mapping set $T = \{\tau_0, \tau_1, ..., \tau_n\}$. Then we use one of the partial mappings as an example. $\tau_0 = \{q_0 \to \mathsf{q}_{10}, q_1 \to \mathsf{q}_{10},$

Algorithm 1: Complementary mapping algorithm based on connectivity between qubits

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Input: \mathcal{AG}_{\mathcal{L}}: The architecture of logical circuit
    \mathcal{AG}_{\mathcal{P}}: The architecture of physical circuit
    T: A partial mapping set obtained by SubgraphMatching
    Output: result: A collection of mapping relations between \mathcal{AG}_{\mathcal{L}} and \mathcal{AG}_{\mathcal{P}}
 1 Initialize result = \emptyset;
 2 l \leftarrow \max_{\tau \in T} \tau.length;
 3 for \tau \in T do
         if l = \tau.length then
              result.add(\tau);
 5
              Q \leftarrow an empty unmapped node queue
 6
              i \leftarrow 1;
 7
              while i \leq \tau.length do
 8
                   Q.push(\{i, i \leq \tau.length \})
 9
                 i \leftarrow i + 1;
10
              while Q is not empty do
11
                    q \leftarrow Q.poll();
12
                    targetAdj \leftarrow \mathcal{AG}_{\mathcal{P}}.adjacencyMatrix();
13
                    queryAdj \leftarrow \mathcal{AG}_{\mathcal{L}}.adjacencyMatrix();
                    cans \leftarrow an empty candidate node list sorted by degree
15
                    cans \cup \{q_m, \ q_m \leq queryAdj[q].length\};
16
                    while cans\ is\ not\ empty\ \mathbf{do}
17
                        q \leftarrow \tau[cans.first];
18
                         k \leftarrow 0;
19
                         cans \leftarrow cans \backslash cans.first;
\mathbf{20}
                         while k < targetAdj[q].length do
21
22
                              if (targetAdj[q][k] \neq -1 \text{ or } targetAdj[k][q] \neq -1)
23
                               and not \tau.contains(k) then
                                  \tau[q] \leftarrow k;
24
                                break;
25
                             k \leftarrow k + 1;
26
                        if k \neq targetAdj[q].length then
27
28
29 return result;
```

-1, $q_2 \to \mathsf{q}_6$, $q_3 \to \mathsf{q}_5$, $q_4 \to \mathsf{q}_{11}$ }, $0 \le i < n$. Here $q_1 \to -1$ means that q_1 is not mapped to any physical node, so we need to perform a mapping completion. In this example, the maximum number of mapped nodes is 4. Next, we demonstrate how τ_0 is completed. We add all unmapped nodes to the queue Q, $Q = \{q_1\}$, and end the loop ends when Q is empty. We put the first element of Q into q, and delete it from Q. Then we get the adjacency matrix of the query graph and the target graph, and traverse the adjacency matrix. We put the nodes q_m adjacent to q into the candidate nodes list cans, which is sorted by the connectivity of q_m and q. We get $cans = \{q_3, q_2, q_4, q_0\}$. Thereafter, we traverse cans and take out of the first element $value = q_3$ in cans, and calculate the phycical node $q = q_5$, $\tau_0(q_3) = q_5$. Finally, we map q to the node connected to q but not yet mapped. If the nodes connected to q have been mapped, the loop continues. In this example, it can be directly mapped to q_0 . In the end, we obtain the mapping

$$\tau_0 = \{q_0 \rightarrow \mathsf{q}_{10}, q_1 \rightarrow \mathsf{q}_0, q_2 \rightarrow \mathsf{q}_6, q_3 \rightarrow \mathsf{q}_5, q_4 \rightarrow \mathsf{q}_{11}\}.$$

5.3 Adjustment

Tabu search The Tabu search algorithm is a type of heuristic algorithm. It uses a tabu list to avoid searching repeated spaces, thereby avoiding deadlock. The algorithm uses amnesty rules to jump out of the local optimum to ensure the diversity of transformed results. The circuit adjustment mainly relies on the Tabu search algorithm, aiming to deal with those large circuits that the current algorithm is difficult to handle and produce an output circuit closer to the optimum solution in a short time.

The following objects are defined in Tabu search: neighborhoods, neighborhood action, tabu list, candidate set, tabu object, evaluation function, and amnesty rule. All the edges that can be swapped in the current map are the neighborhoods. The tabu list avoids local optimum and guarantees the parallelism of auxiliary gates. The tabu object is the object in the tabu list. We try not to use the recently swapped qubits as much as possible, which are added to the tabu list. We perform pruning to reduce search space, because only swaps adjacent to at least one gate node are meaningful. We select the edge in the shortest path that has an intersection with the qubits contained in the gate as the candidate set. The evaluation function selects a SWAP evaluation formula from the candidate set, taking the objective function as the evaluation function in general. The evaluation function satisfies some gates, and the number of SWAP gates added or the depth of the entire circuit should be small. The amnesty rules are used when all objects in the candidate set are banned, or after banning an object, the target value will be greatly reduced.

The calculation of the neighborhoods is shown in Algorithm 2. The input is the current circuit mapping τ_p . The variable qubits contains the mapping of physical qubits to logical qubits, where j=qubits[i] means that the i-th physical qubit has been mapped to the j-th logical qubit. The variable locations represents the mapping of logical qubits to physical qubits, where j=locations[i] means that the i-th logical qubit has been mapped to the j-th physical qubit.

The current layer list of all gates is cl, and the output is a candidate set of the current mapping. The set E contains the edges of all the shortest paths in the physical architecture graph of all the gates in the current layer. Lines 17-37 swap all the edges of candidate set, and calculate the cost of them.

Example 4. Let us consider the mapping

$$\tau_0 = \{q_0 \rightarrow \mathsf{q}_{10}, q_1 \rightarrow \mathsf{q}_0, q_2 \rightarrow \mathsf{q}_6, q_3 \rightarrow \mathsf{q}_5, q_4 \rightarrow \mathsf{q}_{11}\},$$

for $L_0 = \{g_0, g_1\}$, $dist_{cnot}(g_0) = 3$ and $dist_{cnot}(g_1) = 3$. Gate g_1 can be executed directly in the τ_0 mapping, so we delete it from L_0 , but g_0 cannot be executed in the mapping τ_0 . Thus, a circuit adjustment is required. Nodes that cannot be executed join the set $swap_nodes = \{q_0, q_6\}$. The set of shortest paths is $paths = \{\{q_6 \rightarrow q_1 \rightarrow q_0\}, \{q_6 \rightarrow q_5 \rightarrow q_0\}\}$, and then we traverse the shortest paths to calculate the candidate set. The two endpoints of an edge passed by one of the shortest paths should intersect the swap set and join the candidate set. The current candidate set is $\{SWAP(q_6, q_1), SWAP(q_1, q_0), SWAP(q_6, q_5), SWAP(q_5, q_0)\}$.

The circuit mapping algorithm based on Tabu search takes a layered circuit and an initial mapping as input and outputs a circuit that can be executed in the specified architecture graph (see Algorithm 3). The transformed circuit mapping of each layer is used as the initial mapping of the next layer of the circuit. Line 1 regard the initial mapping τ_{ini} as the best mapping τ_{best} . Lines 3-12 cyclically check whether all all gates in the current layer can be executed under the mapping τ_{ini} . If not all the gates are executable or the number of iterations has not reached the given maximum number, the search will continue. Otherwise, the search will terminate. Line 4 gets the current mapping candidate, and Line 7 finds the best mapping in the candidate set. The mapping will first remove the overlapping elements of the candidate set and the tabu list. Then from the remaining candidates, we choose a mapping with the lowest cost. Line 9 are the amnesty rules. When the best candidate is not found, the candidate set elements are all the same as the tabu list elements. The amnesty rules select the mapping with the lowest cost in the candidate set as the best candidate mapping. Lines 10-12 update the best mapping τ_{best} and the current mapping τ_{curr} , and add the SWAP performed by the best mapping to the tabu list tl, indicating that the SWAP has just been performed. The algorithm would try to avoid reswapping the just swapped qubits. Then it will judge whether the termination condition of the algorithm is satisfied. The condition determines whether the number of iterations has reached the maximum number, or the current mapping ensures all gates in the current layer can be executed.

Example 5. Let us continue the previous example. We start searching from the initial mapping. We need to get the candidate SWAP set and select the one with the lower evaluation scores. For $L_0 = \{g_0, g_1\}$, the candidate set is

$$\{SWAP(q_6, q_1), SWAP(q_1, q_0), SWAP(q_6, q_5), SWAP(q_5, q_0)\},\$$

Algorithm 2: Calculate the candidate sets

```
Input: dist: The shortest paths of physical architecture
    qubits: The mapping from physical qubits to logical qubits
    locations: The mapping from logical qubits to physical qubits
    cl: Gates included in the current layer of circuits
    Output: results: The set of candidate solution
 1 Initialize results \leftarrow \emptyset;
 2 E_w \leftarrow \text{Calculate the weight of each edge}
 \mathbf{3} \;\; swap\_nodes \leftarrow \; \text{An empty set of candidate swap nodes}
 4 foreach g \in cl do
         if g is executable then
 5
          cl \leftarrow cl \setminus \{g\};
 6
 7
         else
              q_1 \leftarrow locations[g.control];
 8
              q_{2} \leftarrow locations[g.target];
 9
              swap\_nodes.add(q_t);
10
             swap\_nodes.add(q_2);
11
12 foreach g \in cl do
         q_1 \leftarrow locations[g.control];
13
         q_{\textit{2}} \leftarrow locations[g.target];
14
15
         foreach path \in paths[q_1][q_2] do
              foreach e \in path do
16
                  if \{sour\_node, tar\_node\} \cap swap\_nodes \neq \emptyset then
17
                       new\_qubits \leftarrow qubits;
18
                       new\_locations \leftarrow locations;
19
                        q_1 \leftarrow new\_qubits[e.source];
20
                        q_2 \leftarrow new\_qubits[e.target];
21
                       new\_qubits[e.source] \leftarrow q_2;
\mathbf{22}
23
                       new\_qubits[e.target] \leftarrow q_1;
                       if q_1 \neq -1 then
\mathbf{24}
                        25
                       if q_2 \neq -1 then
26
                        | \begin{array}{c} new\_locations[q_2] \leftarrow q_1; \end{array}
27
                       s \leftarrow \emptyset;
28
                       s.swaps \leftarrow p.swaps \cup \{distance.paths[path_{index}][j]\};
\mathbf{29}
                       s.value \leftarrow compute\_evaluate\_value(dist, new\_locations, cl);
30
                       results \leftarrow results \cup \{s\};
32 return results;
```

Algorithm 3: Tabu search

```
Input: \tau_{ini}: The initial mapping
    tl: Tabu list
    Output: \tau_{best}: The final state and SWAPs
 1 Initialize \tau_{best} \leftarrow \tau_{ini};
 2 iter \leftarrow 1;
                                                                     // Number of iterations
 3 while not musTSAop(iter, \tau_{best}) do
         C \leftarrow \tau_{ini}.candidates();
                                                                                 // candidate set
         if C is empty then
          |break;
  6
         C_{best} \leftarrow find\_best\_candidates(C, tl);
         if C_{best} is empty then
           C_{best} \leftarrow find\_amnesty\_candidates(C, tl);
         \tau_{best} \leftarrow C_{best};
10
         tl \leftarrow tl \cup \{C_{best}.swap\};
11
         iter \leftarrow iter + 1;
12
13 return \tau_{best}
```

and the costs are given as follows.

$$\begin{aligned} & cost(SWAP(\mathsf{q}_6,\mathsf{q}_1)) = 3.0, & & cost(SWAP(\mathsf{q}_1,\mathsf{q}_0)) = 3.0, \\ & cost(SWAP(\mathsf{q}_6,\mathsf{q}_5)) = 3.0, & & cost(SWAP(\mathsf{q}_5,\mathsf{q}_0)) = 3.0. \end{aligned}$$

The algorithm will choose the first SWAP, the mapping becomes

$$\tau_0 = \{q_0 \to \mathsf{q}_{10}, q_1 \to \mathsf{q}_0, q_2 \to \mathsf{q}_1, q_3 \to \mathsf{q}_5, q_4 \to \mathsf{q}_{11}\}.$$

It can be seen that the current mapping ensures the executability of g_0 . The algorithm continues to search for the next layer.

Evaluation functions We can control the search direction by changing the evaluation functions. We test two evaluation functions: one uses the number of auxiliary gates in the generated circuit as the evaluation criterion (1), and the other uses the depth of the generated circuit as the evaluation criterion (2).

$$cost(SWAP(\mathbf{q_m},\mathbf{q_n})) = \sum_{g \in L} (dist[g.control][g.target]) \tag{1}$$

$$cost(SWAP(q_{m}, q_{n})) = Depth(L)$$
(2)

Here $cost(SWAP(q_m,q_n))$ represents the cost of executing all the gates of the current layer L after swapping q_m with q_n . We only calculate the depth between the unmapped gates as in (1) or the distance of the unmapped gates as in (2).

Look ahead We observe that the number of gates in each layer after layering is small. The output of the i-th (i < n) layer is used as the input of the (i+1)-th layer. Note that any swap operation in the i-th layer will affect the mapping of the (i+1)-th layer. If we only consider the gates in the current layer when choosing the swapping gates, the swap only satisfies the requirement of the i-th layer, not necessarily the next layer. Therefore, we take the gates in the (i+x)-th (i+x < n) layer into consideration, where x is the number of forward-looking layers. However, it is necessary to give a higher priority to the execution of the gates in the i-th layer, so we introduce an attenuation factor δ , which controls the influence of the gates in the (i+x)-th layer. Heuristics show that for x=2, $\delta=0.9$, the final effect is approximately the best. Our evaluation functions in (1) and (2) can be adjusted as (3) and (4), respectively.

$$\begin{split} cost(SWAP(\mathbf{q_m},\mathbf{q_n})) &= \sum_{g \in L_i} (dist[g.control][g.target]) + \\ &\delta \times \sum_{j=i}^{i+x} \sum_{g \in L_j} (dist[g.control][g.target]) \end{split} \tag{3}$$

$$cost(SWAP(q_{m}, q_{n})) = Depth(L_{i}) + \delta \times Depth(\sum_{i=i}^{i+x} L_{j}). \tag{4}$$

Complexity Given a logical circuit architecture graph $\mathcal{AG}_{\mathcal{L}} = (V_L, E_L)$ and a physical circuit architecture graph $\mathcal{AG}_{\mathcal{P}} = (V_P, E_P)$, we assume that the initial mapping is τ , the depth of the circuit is d, and the number of qubits is V_L . Tabu search deals with one layer at a time, and searches at most d times. Starting from the initial mapping, we first delete the executable gates of the first layer under the initial mapping. Then, the edges of all the shortest paths of all the gates that are not executable in the current layer are added to the candidate set where at least one node is in the gate mapping. In the worst case, the length of the shortest path is $(|E_P|-1)$ and the size of the candidate set is $(|E_P|-1)$. Each SWAP will make the total distance between the gates smaller. In the worst case, the number of SWAPs is $(|E_P|-1)^{|E_P|-2}$, but our selection strategy will make the number of SWAPs significantly reduced. Our time complexity is $d \times (|E_P|-1)^{(|E_P|-2)}$, and the space complexity is the size of our candidate set (E_P-1) .

6 Experiments

We compare the CSIC algorithm and the circuit adjustment algorithm based on Tabu search TSA with the wghtgraph in [10] and the heuristic algorithm A^* in [25]. All the experiments are conducted on a Windows machine with 3.3GHz CPU and 10G memory.

First, we compare the efficiency of initial mapping on optm [25], CSIC and wghtgraph [10]. In order to observe the results of these two initial mapping

algorithms, we used the same circuit adjustment algorithm A^* [25]. We have tested 159 circuits. Within five minutes, optm, wghtgraph, and CSIC can deal with 121, 106, and 131 circuits, respectively. There are 103 circuits that they can handle. We then compare the wghtgraph algorithm and the CSIC algorithm more closely. The wghtgraph algorithm has 21 circuits with fewer auxiliary gates and 19 circuits with smaller depths, and the CSIC algorithm has 54 circuits with fewer auxiliary gates and 60 circuits with smaller depths. They output 25 circuits with equal depth and 29 circuits with equal auxiliary gates. On average, the auxiliary gates of the CSIC algorithm are reduced by 22.44%, and the depths are reduced by 11.25%.

Next, we compare the optm algorithm and the CSIC algorithm. The optm algorithm has one circuit with fewer auxiliary gates and two circuits with a small depth, while the CSIC algorithm has 99 circuits with fewer auxiliary gates and 98 circuits with a small depth. They have 4 circuits with equal depth and 4 circuits with equal auxiliary gates. The auxiliary gates of the CSIC algorithm are relatively reduced by 27.02%, and the depths are reduced by 14.12%. Table 1 shows the experimental data on 104 circuits. The three initial mapping algorithms are compared according to the depths of the generated circuits using the same A^* algorithm, and the numbers of added auxiliary gates. The column headed by CSIC /optm shows the efficiency improvement of the former upon the latter $(n_{optm} - n_{CSIC})/n_{optm}$.

	optm	wgtgraph	CSIC	CSIC /optm	CSIC /wgtgraph
Depths	168895	163422	145040	14.12%	11.25%
Auxiliary gates	20439	19232	14916	27.02%	22.44%

Table 1. Compare optm, wgtgraph, and CSIC .

We then compare the use of two indicators TSA_{dep} and TSA_{num} that prioritize smaller depths and fewer auxiliary gates, respectively. Using the two indicators as objective functions, we tested 159 circuits. The depths of the final circuits obtained by TSA_{num} are 1.93% smaller than TSA_{dep} on average, and the numbers auxiliary gates added are 4.53% smaller on average. When inserting a SWAP gate, the circuit needs to add 3 CNOT gates, and the depth will be increased by 3. Therefore, if fewer SWAP gates are added, the depths of the circuits will reduce accordingly.

Finally, we compare TSA and wgtgraph. Since the wgtgraph algorithm only uses 2-qubit gates, it is impossible to compare the depths of the generated circuits. Instead, we compare the number of SWAP gates added and the time costs. We set a five-minute timeout period and tested 159 circuits. It turns out that TSA_{num} only takes 461 seconds and TSA_{dep} takes 485 seconds. The wgtgraph algorithm takes 1908 seconds for the 159 circuits, but only produces valid results for 99 circuits, including 64 small circuits, 35 medium circuits, and no circuit output is produced for any of the 44 large circuits. Although Tabu search can

quickly produce results on large circuits, in contrast, more auxiliary gates are added. In the generated circuits obtained by wgtgraph from 99 small and medium circuits, the number of SWAP gates added by wgtgraph is 26.87% (resp. 24.89%) smaller than ${\rm TSA}_{num}$ (resp. ${\rm TSA}_{dep}$) on average. The Tabu search can quickly output converted circuits on large circuits, but wgtgraph cannot get results in a short time. Since our candidate set is too small to be able to process the circuit quickly, but this also leads to an increase in the insertion of additional gates. As to SABRE, when dealing with 159 circuits with a five-minute limit for each circuit, it successfully produces results for only 23 small circuits, 6 medium circuits, and 1 large circuit. The detailed results of the circuit comparisons are in the Appendix.

benchmarks	#circ	TSA_{num}		TSA	TSA_{dep}		wgtgraph		SABRE	
Deficilitatiks	#-CITC.	#succ.	time	#succ.	time	#succ.	time	#succ.	time	
small	66	66	32	66	29	64	587	23	12996	
medium	49	49	45	49	40	35	1183	6	13019	
large	44	44	407	44	432	0	-	0	1340719	
total	159	159	484	159	501	99	-	29	1366734	

Table 2. Compare optm, wghtgraph, SABRE TSA_{num} and TSA_{dep} .

7 Conclusion

We propose a scalable algorithm for quantum circuit conversion. First, we use a combined subgraph isomorphism algorithm and complement the mapping based on the connectivity between qubits to generate a high-quality initial mapping. Secondly we used a look-ahead heuristic search using the influence of the prelayer circuit mapping. Finally, we compared the initial mapping with the latest algorithm and also compared the overall efficiency with wghtgraph, SABRE, and A^* . Experimental results show that the initial mappings generated by CSIC have fewer SWAP gates inserted and the results can be obtained in a short time. Most small and medium circuits can be handled in a few seconds. For large circuits, the results can be obtained within a few minutes, but the cost of insertion may be larger than that of wgtgraph. We introduce a look-ahead method to make each selected SWAP more in line with the constraints of the gates to be processes. In the future, we will investigate how to reduce the number of auxiliary gates inserted and increase the speed. We will also apply the proposed method to more NISQ devices.

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A Experimental details of the SWAP gates added by the output circuits

Circuit	auhit	CNOT	TSA_{num}	TSA_{dep}	optm	wohter	SABRE
name	no.	no.	added	added	added		added
decod24-enable_126	6	149	28	42	60	16	129
4mod5-v0_19	5	16	0	0	0	0	123
4mod5-v0_19 4mod5-v0_18	5	31	$\frac{0}{2}$	5	4	4	_
mod5d2_64	5	25	5	6	8	3	_
4gt4-v0_72	6	$\frac{20}{113}$	14	10	33	14	-
alu-v3_35	5	18	2	4	8	2	_
4gt4-v0_73	6	179	27	34	76	12	_
alu-v3_34	5	24	2	3	7	2	_
3_17_13	3	17	0	0	6	0	0
4gt4-v0_78	6	109	12	8	48	4	124
4gt4-v0_79	6	105	17	17	48	3	-
4mod7-v1_96	5	72	16	19	27	7	_
mod10_171	5	108	17	20	39	9	_
ex2_227	7	275	48	59	121	33	_
mod10_176	5	78	14	14	38	8	_
0410184_169	5	9	2	2	49	3	_
4mod5-v0_20	5	10	0	0	4	0	_
aj-e11_165	5	69	8	8	33	7	63
alu-v1_28	5	18	$\frac{\circ}{2}$	4	11	2	-
4gt12-v0_86	6	116	28	33	48	3	_
4gt12-v0_87	6	112	$\frac{1}{27}$	32	45	2	_
4gt12-v0_88	6	86	5	5	25	4	_
alu-v1_29	5	17	4	4	11	2	13
ham7_104	7	149	28	34	68	12	_
C17_204	7	205	26	53	99	22	-
xor5_254	6	5	0	0	1	0	3
hwb4_49	5	107	14	15	38	11	157
rd73_140	10	104	23	26	35	20	-
$decod24-v0_38$	4	23	0	0	6	0	-
rd53_131	7	200	39	39	98	24	-
rd53_133	7	256	37	47	102	27	160
rd53_135	7	134	28	29	38	23	-
$decod24-v2_43$	4	22	0	0	9	0	39
rd53_138	8	60	14	16	23	9	-
rd32-v0_66	4	16	0	0	6	0	-
4gt13-v1_93	5	30	0	0	13	0	-
graycode6_47	6	5	0	0	0	0	-
$4 \bmod 5 - bdd_2 287$	7	31	3	6	8	6	34
ham3_102	3	11	0	0	3	0	12
4gt4-v0_80	6	79	5	5	22	5	-
ex-1_166	3	9	0	0	3	0	-
$mod5mils_65$	5	16	0	0	6	0	-
0example	5	9	1	2	3	3	-
alu-v $4_{-}36$	5	51	12	8	22	4	41
alu-v4 $_{-}37$	5	18	2	4	8	2	-
ex1_226	6	5	0	0	1	0	15
one-two-three-v0_98	5	65	11	13	32	10	-
one-two-three-v0_97	5	128	23	23	64	16	-
one-two-three-v3_101	5	32	3	4	14	3	-
rd32_270	5	36	3	3	6	6	24

 ${\bf Table~3.}$ Comparison of the number of SWAP gates added by the output circuit on the IBM Q20

C::t	1-:4	CNOT	TTC A	TTC A	4	1	CADDE
Circuit	_		TSA_{num}	TSA_{dep}	optm		SABRE
name	no.	no.	added	added	added		added
rd53_130	7	448	89	100	190	49	-
rd53_251	8	564	104	131	230	45	-
4mod5-v1_24	5	16	0	0	3	0	-
mod5adder_127	6	239	21	56	111	20	-
4_49_16	5	99	20	17	40	10	-
hwb5_53	6	598	141	168	173	59	-
ex3_229	6	175	10	9	50	11	-
4gt10-v1_81	5	66	14	15	28	6	-
alu-v2_32	5	72	15	17	27	7	-
alu-v2_31	5	198	42	54	85	13	-
alu-v2_30	6	223	41	45	96	20	-
sf_276	6	336	12	52	138	12	-
$decod24-v1_41$	5	38	4	4	14	3	28
sf_274	6	336	34	21	82	12	-
4gt4-v1_74	6	119	17	24	37	9	-
alu-v2_33	5	17	4	4	8	2	-
$cnt3-5_{-}179$	16	85	6	6	35	4	-
$4 \mod 5 - v1_2 $	5	11	0	0	5	0	0
$4 \mod 5 - v1_2 3$	5	32	5	5	4	3	55
mini_alu_305	10	77	10	20	28	8	123
alu-v0_26	5	38	7	10	13	3	-
alu-bdd_288	7	38	4	12	16	6	-
alu-v0_27	5	17	2	4	11	2	9
4gt13_91	5	49	7	7	10	2	-
$4gt5_{-}77$	5	58	12	12	20	6	-
4gt13_92	5	30	0	0	14	0	-
4gt5_76	5	46	7	10	24	5	39
4gt5_75	5	38	5	12	16	4	53
4gt12-v1_89	6	100	11	21	38	4	-
one-two-three-v1_99	5	59	12	10	26	7	-
4gt13_90	5	53	7	7	13	3	-
ising_model_10	10	90	0	0	5	0	-
4gt11_84	5	9	0	0	3	0	12
4gt11_83	5	14	0	0	0	0	-
mod5d1_63	5	13	0	0	1	0	-
4gt11_82	5	18	1	1	1	1	-
decod24-v3_45	5	64	15	15	32	8	64
rd32-v1_68	4	16	0	0	6	0	24
mini-alu_167	5	126	27	27	49	11	-
one-two-three-v2_100	5	32	3	4	8	3	_
4mod7-v0_94	5	72	8	13	36	9	-
cm82a_208	8	283	41	69	84	33	-
mod8-10_178	6	152	5	20	13	7	-
mod8-10_177	6	196	14	33	58	13	230
majority_239	7	267	39	43	105	33	_
miller_11	3	23	0	0	9	0	36
decod24-bdd_294	6	32	4	4	9	4	36
total	551	9244	1372	1738	3481	800	-
Count	551	0211	1012	1.00	0101	000	

 ${\bf Table~4.}$ Comparison of the number of SWAP gates added by the output circuit on the IBM Q20

Circuit	aubit	CNOT	TSA_{num}	TSA_{dep}	optm	wehter	SABRE
name	no.	no.	added	added	added	added	added
max46_240	10	11844	3473	4545	-	-	-
rd73_252	10	2319	586	761	_	_	_
cycle10_2_110	12	2648	919	1216	961	_	_
sgrt8_260	12	1314	379	492	457	_	_
urf4_187	11	224028	54785	60140	-	_	_
sqn_258	10	4459	1199	1420	_	_	_
f2_232	8	525	87	124	218	_	_
radd_250	13	1405	386	489	511	_	_
ham15_107	15	3858	1326	1689	-	_	_
sao2_257	14	16864	5346	7178		_	_
sym9_148	10	9408	1865	2432	_	_	_
urf5_280	9	23764	6989	8730	_	_	_
square_root_7	15	3089	812	2150	_	_	-
sys6-v0_111	10	98	23	26	38	_	-
hwb7_59	8	10681	$\frac{25}{2687}$	3551	3722	_	-
	$\frac{\circ}{12}$	148	38	55	54	_	248
sym9_146	11		93		147	_	240
$\begin{array}{c} \text{wim}_266\\ \text{urf}2_152 \end{array}$	8	$427 \\ 35210$		120 11921		-	-
urf5_152 urf5_159	9	71932	9181 20258	25505	10577	-	-
urf2_277	8				3782	-	-
life_238	11	10066	$2807 \\ 2762$	3798 3576	3162	-	-
		9800			-	-	-
root_255	13	7493	2128	3035	-	-	-
9symml_195	11	15232	4553	5986	-	-	-
sym10_262	12	28084	8534	11033	971	-	-
dc1_220	11	833	226	207	371	-	-
cm42a_207	14	771	182	229	294	-	-
rd53_311	13	124	26	48	47	-	-
dc2_222	15	4131	1383	1773	-	-	-
rd84_142	15	154	49	58	50	-	-
sym6_145	7	1701	317	449	750	-	-
co14_215	15	7840	3078	3819	-	-	-
cnt3-5_180	16	215	59	74	79	-	-
cm152a_212	12	532	103	129	168	-	-
sym6_316	14	123	30	39	56	-	-
mlp4_245	16	8232	2780	3490	-	-	-
hwb8_113	9	30372	10749	16489	-	-	-
qft_16	16	240	90	147	-	-	-
plus63mod4096_163	13	56329	19759	24273	-	-	-
urf1_149	9	80878	22551	28516	-	-	-
urf3_155	10	185276	50842	62903	-	-	-
urf3_279	10	60380	17999	23318	-	-	-
hwb9_119	10	90955	22946	30031	-	-	-
plus63mod8192_164	14	81865	28022	36207	-	-	-
pm1_249	14	771	182	229	294	-	-
sym9_193	11	15232	4382	5518	-	-	-
misex1_241	15	2100	480	754	600	-	-
urf1_278	9	26692	8010	10217	-	-	-
squar5_261	13	869	219	313	290	-	-
ground_state_estimation_10	1	154209	11671	22886	-	-	-
adr4_197	13	1498	516	670	-	-	-

 ${\bf Table~5.}$ Comparison of the number of SWAP gates added by the output circuit on the IBM Q20

F. Author et al.

Circuit	qubit	CNOT	TSA_{num}	TSA_{dep}	optm	wghtgr	SABRE
name	no.	no.	added	added	added	added	added
hwb6_56	7	2952	698	933	909	-	-
clip_206	14	14772	5430	6865	-	-	-
cm85a_209	14	4986	2088	2225	-	-	-
$rd84_{-}253$	12	5960	1849	2333	-	-	-
$dist_223$	13	16624	5623	7431	-	-	_
inc_237	16	4636	1193	1667	-	-	_
$qft_{-}10$	10	90	23	34	30	-	-
urf6_160	15	75180	27524	32452	-	-	-
con1_216	9	415	86	118	177	-	-

Table 6. Comparison of the number of SWAP gates added by the output circuit on the IBM $\mathbf{Q}20$

B Experimental details of the depths of the output circuits

Circuit	qubit	CNOT	depths	TSA_{num}	TSA_{dep}	optm
name	no.	no.	no.	depths	depths	depths
decod24-enable_126	6	149	190	233	275	470
4mod5-v0_19	5	16	21	16	16	21
4mod5-v0_18	5	31	40	37	46	54
mod5d2_64	5	25	32	40	43	67
4gt4-v0_72	6	113	137	155	143	297
alu-v3_35	5	18	22	24	30	60
4gt4-v0_73	6	179	227	260	281	586
alu-v3_34	5	24	30	30	33	63
3_17_13	3	17	22	17	17	52
4gt4-v0_78	6	109	137	145	133	352
4gt4-v0_79	6	105	132	156	156	345
4mod7-v1_96	5	72	94	120	129	218
mod10_171	5	108	139	159	168	335
ex2_227	7	275	355	419	452	899
mod10_176	5	78	101	120	120	104
cycle10_2_110	12	2648	3386	5405	6296	7467
0410184_169	5	9	6	15	15	253
4mod5-v0_20	5	10	12	10	10	32
sqrt8_260	12	1314	1661	2451	2790	3561
aj-e11_165	5	69	86	93	93	250
alu-v1_28	5	18	22	24	30	70
f2_232	8	525	668	786	897	1672
$radd_250$	13	1405	1781	2563	2872	3985
4gt12-v0_86	6	116	135	200	215	334
4gt12-v0_87	6	112	131	193	208	324
4gt12-v0_88	6	86	108	101	101	222
alu-v1_29	5	17	22	29	29	64
ham7_104	7	149	185	233	251	491
C17_204	7	205	253	283	364	688
xor5_254	6	5	5	5	5	10
hwb4_49	5	107	134	149	152	308
rd73_140	10	104	92	173	182	185
$decod24-v0_{-}38$	4	23	30	23	23	61
rd53_131	7	200	261	317	317	677
rd53_133	7	256	327	367	397	777
rd53_135	7	134	159	218	221	331
sys6-v0_111	10	98	75	167	176	188
$decod24-v2_43$	4	22	30	22	22	75
hwb7_59	8	10681	13437	18742	21334	29601
rd53_138	8	60	56	102	108	114
rd32-v0_66	4	16	20	16	16	51
sym9_146	12	148	127	262	313	309
4gt13-v1_93	5	30	39	30	30	102
graycode6_47	6	5	5	5	5	5
wim_266	11	427	514	706	787	1180
$urf2_{-}152$	8	35210	44100	62753	70973	90299
urf2_277	8	10066	11390	18487	21460	26548
4mod5-bdd_287	7	31	41	40	49	71
ham3_102	3	11	13	11	11	28
4gt4-v0_80	6	79	101	94	94	206

Table 7. Comparison of the depth of the output circuit on the IBM Q20

Circuit	auhit	CNOT	dontha	TSA_{num}	TSA_{dep}	ontm
	_	l	_			optm
name	no.	no.	no.	depths 9	depths 9	depths
ex-1_166 mod5mils_65	3 5	9 16	21	16	9 16	28 52
	5	9	$\frac{21}{6}$	12	16 15	15
0example alu-v4_36	5	51	66	87	75	170
alu-v4_36 alu-v4_37	5	18	$\frac{66}{22}$	24	30	60
ex1_226	$\begin{array}{c c} 5 \\ 6 \end{array}$	5	$\frac{22}{5}$	5	50 5	10
one-two-three-v0_98	5	65	82	98	$\frac{3}{104}$	$\begin{vmatrix} 10 \\ 234 \end{vmatrix}$
one-two-three-v0_97	5	128	163	197	$\frac{104}{197}$	443
one-two-three-v3_101	5	32	40	41	44	95
rd32_270	5	36	$\frac{40}{47}$	45	44 45	76
dc1_220	11	833	1041	1511	$\frac{45}{1454}$	2711
rd53_130	7	448	569	715	$\frac{1434}{748}$	1417
rd53_251	8	564	712	876	957	1767
cm42a_207	14	771	940	1317	$\frac{957}{1458}$	2279
rd53_311	13	124	130	202	$\frac{1458}{268}$	300
4mod5-v1_24	5	16	21	16	208 16	36
mod5adder_127	$\begin{array}{ c c c c c c c c c c c c c c c c c c c$	$\frac{10}{239}$	$\frac{21}{302}$	302	$\frac{10}{407}$	817
4_49_16	5	239 99	$\frac{302}{125}$	159	$\frac{407}{150}$	320
4_49_10 hwb5_53	$\begin{array}{c c} 5 \\ 6 \end{array}$	598	758	1021		1560
ex3_229	6	175	226	205	$\frac{1102}{202}$	462
	15	154		301	$\frac{202}{328}$	253
rd84_142	1	1	110			
4gt10-v1_81	5	66	84	108	111	210
alu-v2_32	5 5	72	92	117	123	215
alu-v2_31 alu-v2_30	6	198 223	$\frac{255}{285}$	324 346	$\frac{360}{358}$	650 734
sym6_145 sf_276	7	1701	2187	2652	3048	5716
	6	336	435	372	492	1096
decod24-v1_41	5 6	38	50	50	50	120
sf_274	6	336	436 154	438 170	399	822 329
4gt4-v1_74		119	$\frac{154}{22}$		191	
alu-v2_33	5	17		29	29 427	59
cnt3-5_180 cm152a_212	16 12	215 532	209 684	392 841	$437 \\ 919$	482 1423
cnt3-5_179	16	85	61	103	103	166
		l				
sym6_316 4mod5-v1_22	14 5	123 11	$\frac{135}{12}$	213 11	$\frac{240}{11}$	378 37
4mod5-v1_22 4mod5-v1_23	5	32	41	47	$\frac{11}{47}$	55
mini_alu_305	$\frac{3}{10}$	32 77	71	107	137	187
alu-v0_26	l		49	59	68	108
alu-bdd_288	$\begin{array}{ c c c }\hline 5 \\ 7 \end{array}$	38		l		
	5	38	48	50	74	112
alu-v0_27 4gt13_91	5	17	21 61	23	29	63
		49		70	70	108
4gt5_77	5	58	74	94	94	170
4gt13_92	5	30	38	30 67	30 76	103
4gt5_76	5	46	56	67	76 74	171
4gt5_75	5	38	47	53	74	127
4gt12-v1_89	6 5	100	130	133	163	313
one-two-three-v1_99	1	59	76	95	89 74	194
4gt13_90	5	53	65	74	74	124
pm1_249	14	771	940	1317	1458	2279

Table 8. Comparison of the depth of the output circuit on the IBM $\mathrm{Q}20$

Circuit	qubit	CNOT	depths	TSA_{num}	TSA_{dep}	optm
name	no.	no.	no.	depths	depths	depths
ising_model_10	10	90	52	90	90	107
misex1_241	15	2100	2676	3540	4362	5326
4gt11_84	5	9	11	9	9	25
4gt11_83	5	14	16	14	14	16
$mod5d1_63$	5	13	13	13	13	17
4gt11_82	5	18	20	21	21	25
squar5_261	13	869	1051	1526	1808	2309
$decod24-v3_45$	5	64	84	109	109	244
rd32-v1_68	4	16	21	16	16	52
hwb6_56	7	2952	3736	5046	5751	7773
mini-alu_167	5	126	162	207	207	400
one-two-three-v2_100	5	32	40	41	44	80
4mod7-v0_94	5	72	92	96	111	270
cm82a_208	8	283	340	406	490	699
mod8-10_178	6	152	193	167	212	243
mod8-10_177	6	196	251	238	295	525
majority_239	7	267	344	384	396	839
$qft_{-}10$	10	90	37	159	192	135
miller_11	3	23	29	23	23	75
decod24-bdd_294	6	32	40	44	44	86
con1_216	9	415	508	673	769	1197
total	823	83416	103023	145372	164848	224731

Table 9. Comparison of the depth of the output circuit on the IBM $\mathrm{Q}20$

name 1 max46_240 rd73_252	no. 10 10	CNOT no. 11844 2319	no.	$\frac{\text{TSA}_{num}}{\text{depths}}$ $\frac{22263}{}$	TSA_{dep} depths	optm depths
max46_240 rd73_252	10 10 11	11844 2319	14257	_	depths	depths
rd73_252	10 11	2319		าาาคา		
	11				25479	-
urf4_187			2867	4077	4602	-
I I	10		264330	388383	404448	-
1		4459	5458	8056	8719	-
1	15	3858	4819	7836	8925	-
	14	16864	19563	32902	38398	-
-3	10	9408	12087	15003	16704	-
urf5_280	9	23764	27822	44731	49954	-
square_root_7	15	3089	3847	5525	9539	-
urf5_159	9	71932	89148	132706	148447	-
life_238	11	9800	12511	18086	20528	-
root_255	13	7493	8839	13877	16598	-
$9 \text{symml}_{-} 195$	11	15232	19235	28891	33190	-
sym10_262	12	28084	35572	53686	61183	-
$dc2_222$	15	4131	5242	8280	9450	-
co14_215	15	7840	8570	17074	19297	-
mlp4_245	16	8232	10328	16572	18702	-
hwb8_113	9	30372	38717	62619	79839	-
$qft_{-}16$	16	240	61	510	681	-
plus63mod4096_163	13	56329	72246	115606	129148	-
urf1_149	9	80878	99586	148531	166426	-
urf3_155	10	185276	229365	337802	373985	-
urf3_279	10	60380	70702	114377	130334	-
hwb9_119	10	90955	116199	159793	181048	-
plus63mod8192_164	14	81865	105142	165931	190486	-
$sym9_{-}193$	11	15232	19235	28378	31786	-
ising_model_13	13	120	46	120	120	-
urf1_278	9	26692	30955	50722	57343	-
ising_model_16	16	150	57	150	150	-
ground_state_estimation_10	13	154209	217236	189222	222867	-
$adr4_{-}197$	13	1498	1839	3046	3508	-
clip_206	14	14772	17879	31062	35367	-
cm85a_209	14	4986	6374	11250	11661	-
rd84_253	12	5960	7261	11507	12959	_
dist_223	13	16624	19694	33493	38917	-
inc_237	16	4636	5864	8215	9637	-
urf6_160	15	75180	93645	157752	172536	-

Table 10. Comparison of the depth of the output circuit on the IBM $\mathrm{Q}20$