

# Automated Window-Based Partitioning of Quantum Circuits

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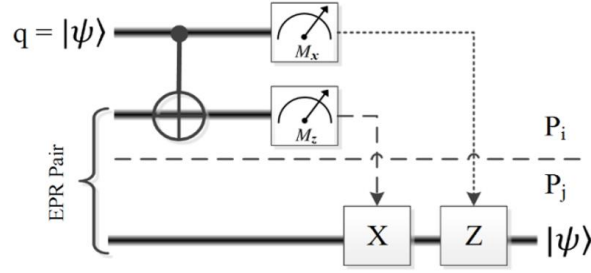
**Abstract.** Developing a scalable quantum computer as a single processing unit is challenging due to technology limitations. A solution to deal with this challenge is distributed quantum computing where several distant quantum processing units are used to perform the computation. The main design issue of this approach is costly communication between the processing units. Focused on this issue, in this paper, an efficient partitioning approach is proposed which combines both gate and qubit teleportation concepts in an efficient manner to minimize the communication. Experimental results show the proposed approach on average reduces the communication cost by about 29.5% in comparison with the best approaches in the literature.

*Keywords:* Quantum Circuit, Distributed Quantum Computing, Window-based Partitioning

## 1. Introduction

While feature size in VLSI technology enters 7 nm and beyond, quantum effects should be handled [1]. However, managing quantum effects in a controlled manner may also be utilized as a feature. Feynman originally suggested [2] using these effects to efficiently solve problems that are intractable on classical computers and called the device a quantum computer. The circuit model of quantum computation is similar to the circuit model consisting of a discrete set of gates found in conventional computing. Sequences of one- and two-qubit operations constitute the fundamental logic for evolving a quantum state. However, there are some unique characteristics for quantum computing such as superposition, entanglement, and the inability to copy arbitrary quantum states [3].

Although many challenges hinder the realization of a practical quantum system, we believe that the design space for a future quantum computer should be explored now that it helps to organize the plethora of proposed quantum technologies, fault tolerance



**Figure 1.** Teleportation circuit to transmit qubit  $q$  from partition  $P_i$  to  $P_j$  using a single EPR pair shared between the partitions  $P_i$  and  $P_j$

methods, and other realization choices. However, when both hardware and architecture parameters are considered, the design space grows significantly. Therefore, to manage the design space complexity, a CAD flow is needed to streamline the design process and to enable us to design large quantum circuits.

A homogeneous organization may be acceptable for a small-scale quantum computer but to build a practical full-scale quantum system, distributed computation with the necessary communications is required [4]. The distributed quantum architectures [4, 5, 6, 7, 8] are organized as quantum processing units (QPU) connected by some interfaces such as photonic networks. In such architectures, the communications have considerably more latency and are more error-prone than local operations [4, 5]. For example, in the MUSIQC hardware proposed in [5], the communication latency between QPUs is in the order of milliseconds while other operations are in the order of microseconds. Thus, minimizing the communications between QPUs has a significant effect on the final latency. The communications between QPUs can dramatically decrease if qubits of a circuit are effectively partitioned into QPUs. Focused on this issue, in this paper, we propose a partitioning approach based on a windowing strategy to distribute qubits among QPUs in a manner that the communication cost is minimized.

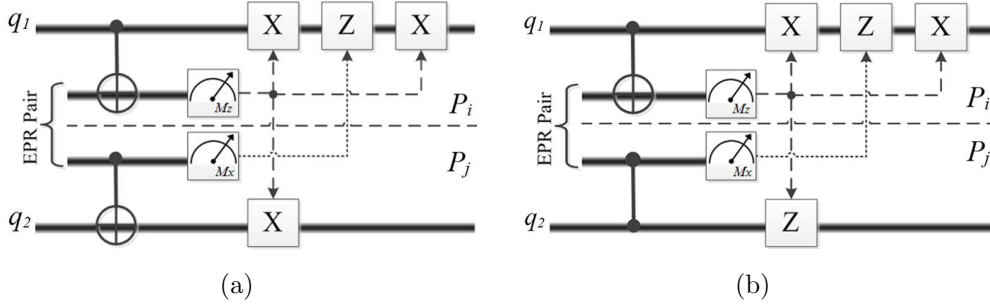
The remainder of this paper is organized as follows: Section 2 contains some basic concepts in the field. An overview of the prior work is presented in Section 3. Section 4 discusses the proposed approach in detail. Experimental results are discussed in the Section 5, and finally Section 6 concludes the paper.

## 2. Background

In this section, some terminologies and concepts are explained that would help to give a better understanding of the proposed approach.

Using teleportation to transmit data qubits from a source to a destination is known as data teleportation or teledata [9]. This procedure can be summarized in Figure 1. Performing a multi-qubit gate using teleportation, without placing the target qubits next to each other, is known as gate teleportation or telegate [10]. Figure 2 shows the circuits to perform CNOT and CZ gates remotely, based on the telegate mechanism.

The teledata and telegate concepts lead to two partitioning approaches namely



**Figure 2.** Telegate-based application of a) CNOT and b) CZ gates on qubits  $q_1$  and  $q_2$  using an EPR pair shared between partitions  $P_i$  and  $P_j$

*gate partitioning* and *qubit partitioning*, respectively. In the gate partitioning, a quantum circuit is partitioned according to its gates. In this approach, it is decided to which partition (QPU) a gate must be assigned. If the qubits of that gate are not in the chosen partition, they are transmitted into that using teledata. On the other hand, the qubit partitioning approach partitions the qubits of a circuit and decides where each qubit should be placed. For applying a multi-qubit gate, if the involved qubits are in different partitions, the gate will be applied remotely via telegate; otherwise, the gate can be performed locally.

### 3. Related Work

Quantum circuit design flow like its classical counterpart can be partitioned into two main processes: synthesis and physical design. The physical design process maps the gate-level netlist generated from the synthesis process onto a physical layout. Several studies have been done on automation of different steps of the physical design process. Some researchers [6, 7, 11, 12, 13, 14, 15, 16] worked on the entire physical design flow and proposed techniques for each of its step while others [17, 18, 19, 20] proposed some techniques for scheduling of a quantum circuit on a layout. Mohammadzadeh et al. [21, 22, 23] introduced the physical synthesis concept for quantum circuits and proposed some practical physical synthesis techniques [21, 23, 24, 25]. Since the main focus of this paper is on the partitioning step, in the rest of this section, the partitioning techniques are reviewed in more detail.

Squash 2 [26] partitions the quantum circuits based on the gate partitioning approach by utilizing METIS [27] as the partitioning tool. Moghadam et al. [28] apply a min-cut placement-aware partitioning approach [29] to divide a quantum dataflow graph of a circuit into smaller manageable parts. Wang et al. [30] modified a graph partitioning algorithm presented in [31] to find the minimum cut of the qubit interaction graph. Ahsan et al. [7, 32] used an efficient graph-theoretic algorithm [33] to assign qubits to QPUs. They first generate the adjacency matrix  $P$  of an  $N$ -qubit circuit where  $P[i][j]$  is the total number of interactions between qubits  $q_i$  and  $q_j$ . Then  $P$  is converted

into its corresponding Laplacian matrix as below:

$$L[i][j] = \begin{cases} \sum_{k=1}^N P[i][k] & i = j \\ -P[i][j] & o.w. \end{cases}$$

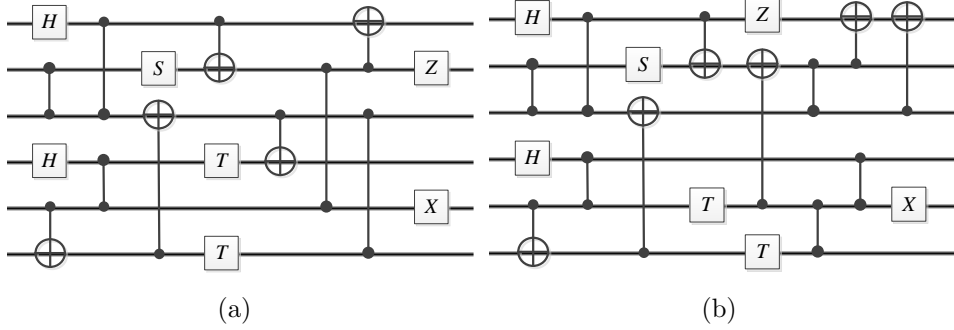
Eigenvalues of  $L$  are computed and the eigenvector  $V_2$  corresponding to the second smallest eigenvalue is selected. Sorting  $V_2$  can determine the best order of qubits in a line in such a way that the weighted sum of the distances between the qubits is minimized. As the last step, this line of qubits is broken into some partitions and assigned to the QPUs.

Mohammadzadeh and Sargaran [6] proposed SAQIP architecture and used the multilevel k-way hypergraph partitioning algorithm introduced in [34] to partition the qubits into QPUs. In [35], the authors call METIS [27] iteratively to separate the qubits and place them on a 2D nearest-neighbor architecture. Zomorodi-Moghadam et al. [36] proposed a gate partitioning procedure to minimize the number of teleportation operations. In that work, an additional exhaustive search is applied to decide how each two-qubit quantum gate should be implemented. This increases the runtime exponentially in the number of partitions and gates, making it futile in practice. However, in a recent work, they reduced the complexity of their method by proposing a genetic algorithm to solve the partitioning problem more efficiently [37].

There are some approaches proposed for quantum physical design automation on single-processor but topologically-constrained architectures. Childs et al. [38] used the token swapping framework [39] and a 4-approximation algorithm [40] to insert a minimal sequence of SWAP gates into the circuit and transform an input quantum circuit to a hardware-compliant one. Chakrabarti et al. [41] proposed a balanced graph partitioning technique to find global ordering of qubit lines to achieve the Linear-Nearest-Neighbor architecture with minimum number of SWAP gates by using pmetis [42], an existing multilevel graph partitioning tool. Minimum linear arrangement problem [43] employed in [44] tries to insert minimum number of SWAP gates in different parts of an interaction graph. A novel reverse traversal technique was proposed in [45] to choose the initial mapping with the consideration of the whole circuit. It takes the following gates and previous mappings into account to reduce the overhead of 2-qubit gates and movements. The authors of [46] proposed an efficient heuristic method for logical to physical qubit mapping for linear devices. This has been realized by transforming the mapping problem into an undirected graphical representation and then has implemented spectral graph theory-based approach for placing logical qubits. All these approaches focus on mapping a given circuit on a topologically-constrained architectures while our architecture is a distributed one with less constraints.

#### 4. Our Proposed Approach: Window-based Quantum Circuit Partitioning (WQCP)

The main drawback of *qubit partitioning* approaches is that they convert a circuit into an untimed qubit interaction graph and try to partition it. Although this assignment



**Figure 3.** Gate partitioning (qubit partitioning) leads to less communication cost than qubit partitioning (gate partitioning) for the circuit a (b).

is done in a manner that more interacting qubits are attempted to be assigned to the same partition, not using timing information in these approaches makes them inefficient. In other words, since qubits may interact in different parts of a circuit, the partitioning methods that ignore timing information do not generate good results. Therefore, a better solution is to partition the circuit based on the information in local connectivity patterns and to change partitions of qubits if the connectivity pattern of the qubits is changed while the circuit proceeds. On the other hand, the main drawback of *gate partitioning* methods is that they force the qubits of multi-qubit gates to be transported to the same partition to apply the gate. However, allowing remote application of multi-qubit gates can mitigate unnecessary forward and backward transfers.

For example, in Figure 3, the goal is partitioning of the circuits into two parts, each consisting of three qubits. The optimal communication costs of the circuit in Figure 3-a using gate and qubit partitioning approaches are 2 and 4, respectively. On the other hand, those achieved for the circuit in Figure 3-b using gate and qubit partitioning approaches are 4 and 2, respectively. Therefore, gate partitioning generates a better result than qubit partitioning for the circuit of Figure 3-a while for the circuit of Figure 3-b, qubit partitioning outperforms gate partitioning. Therefore, the superiority of a method over the other depends on the interaction pattern of the qubits of the circuit. Considering this observation, it seems that combining two partitioning approaches can improve the communication cost by mitigating the drawbacks of existing approaches.

Focusing on this issue, we propose a hybrid partitioning approach, called WQCP, which combines both *telegate* and *teledata* ideas in an efficient manner to minimize the communication cost. The pseudo-code of the proposed algorithm is given in Algorithm 1. In the first step, single-qubit gates are removed from the circuit because single-qubit gates can be applied without any communication regardless of the partition the target qubit is assigned to. Then, the resulting circuit is leveled and a weighted window with the length of  $L_W$  is moved along the circuit from the first level to the last one, level by level. Let  $L_C$ ,  $G_L$ , and  $C_L$  be the number of levels of the circuit, the set of gates in level  $L$  and the sub-circuit contained in the window of length  $L_W$  beginning at

**Algorithm 1** WQCP**Input:** A quantum circuit ( $C_{in}$ ), Window length ( $L_W$ )**Output:** The partitioned circuit, Total communication cost (Number of teledatas and telegates)

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1:  $C = rmvSingleQubitGates(C_{in});$  //Remove single-qubit gates from the circuit  $C_{in}$ 
2:  $L_C = levelize(C);$  //Levelize the circuit  $C$  and return the total number of levels of  $C$ 
3:  $nTD = 0;$  //Initialize the number of teledatas with zero
4:  $nTG = 0;$  //Initialize the number of telegates with zero
5: for each  $L \in Levels = \{1, 2, \dots, L_C\}$  do
6:    $C_L = getWindow(C, L, L_w);$  //Get the sub-circuit surrounded by the window with length  $L_W$ 
   started from level  $L$ 
7:    $P_L = subPartitioning(C_L);$  //Partition  $C_L$  based on qubit partitioning approach
8:    $nTD+ = countTG(P_L, P_{L-1});$  //Return the number of teledatas by comparing  $P_L$  with  $P_{L-1}$ 
9:    $G_L = getGatesAtLevel(L);$  //Return the gates of level  $L$ 
10:   $nTG+ = countTD(G_L, P_L);$  //Return the number of gates of  $G_L$  which must be applied
   remotely by telegate
11: end for

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$L$ , respectively. For each level  $L$ ,  $1 \leq L \leq L_C$ ,  $C_L$  is partitioned based on the qubit partitioning approach. This operation is denoted by  $subPartitioning(C_L)$ . By this, the partition of each qubit is determined for applying the gates of  $G_L$  and is denoted by  $P_L$ . For each gate  $g \in G_L$ , if its qubits are placed in different partitions, the gate will be applied remotely by telegate. Otherwise,  $g$  is applied locally. For two successive levels,  $L - 1$  and  $L$ , if  $subPartitioning(C_L)$  changes the partition of one qubit with respect to its previous partition which is determined by  $subPartitioning(C_{L-1})$ , that qubit is transported to the new partition using teledata. In the rest of this section, the subPartitioning algorithm is explained followed by an example.

#### 4.1. $subPartitioning(C_L)$ algorithm

The subPartitioning function implements a min-cut partitioning algorithm [27, 47, 48] which takes a sub-circuit  $C_L$  as input and partitions its qubits. To do so, the sub-circuit  $C_L$  is modeled using a weighted graph whose vertices are qubits and its edges between two vertices are weighted according to the number of gates applied to the corresponding qubits. The weights of the edges are calculated as follows. Interactions between qubits in different levels of  $C_L$  have different importance in our approach. This is because  $subPartitioning(C_L)$  determines the partition of each qubit only for applying the gates of  $G_L$ , and  $subPartitioning(C_{L+1})$  may change the location of qubits for the gates of the next level. Therefore, a weighted window is used to reflect this difference in the importance of interactions between the qubits in different levels. To this end, the weight of a window is defined as  $W = \{w_k\}$ , where  $1 \leq k \leq L_W$  and  $w_k$  represents the importance of the interactions between the qubits in the  $k^{th}$  level of the sub-circuit  $C_L$ . The weight of each edge between two vertices  $q_i$  and  $q_j$ , denoted by  $E(q_i, q_j)$ , is defined as the weighted sum of interactions between qubits  $q_i$  and  $q_j$  in  $C_L$ . This weight can be

**Table 1.** Input parameters and variables of our ILP model

Type	Symbol	Description
NUCC	$N$	Number of qubits in the sub-circuit $C_L$
	$M$	Total number of partitions
	$capacity$	Maximum number of qubits which can be assigned to each partition simultaneously
	$w_p$	Determines how much qubits tend to stay in their previous partitions
	$adjMat[N][N]$	An $N \times N$ matrix where $adjMat[i][j]$ is equal to $E(q_i, q_j)$ according to Eq. 1
	$prevParts[M][N]$	An $M \times N$ binary matrix where $prevParts[p][i]$ is equal to 1 if and only if the previous partition of qubit $q_i$ is partition $p$ (Previous partition of a qubit is determined by $subPartitioning(C_{L-1})$ ).

formulated as:

$$E(q_i, q_j) = \sum_{k=1}^{L_W} g_{ij}^k \times w_k \quad (1)$$

where  $g_{ij}^k$  is zero if there is not any gate applied to qubits  $q_i$  and  $q_j$  in the  $k^{th}$  level of the sub-circuit  $C_L$  and otherwise it is equal to the number of needed EPR pairs to apply the gate by telegate.

In addition to the interactions of the qubits in  $C_L$ ,  $subPartitioning(C_L)$  should consider the previous partition of each qubit, which is determined by  $subPartitioning(C_{L-1})$ . For doing so, a dummy vertex  $p_i$  is added to the graph corresponding to each partition  $i$  and each qubit is connected to its previous partition vertex with weight  $w_p$ , where  $w_p$  represents how much a qubit tends to stay in its previous partition. We formulate  $subPartitioning(C_L)$  as an ILP problem. The input parameters and variables of the ILP model are given in Table 1.

Our model minimizes the following cost function:

$$\sum_{i=1}^N \sum_{j=i}^N cut[i][j] \times adjMat[i][j] + \sum_{n=1}^N w_p \times migrate[n]$$

It consists of two terms. The first term is the weighted sum of cuts and the second term is the cost incurred by migrating qubits from their previous partitions.

The constraints of the ILP model are listed below:

- Total number of qubits assigned to each partition must not exceed its capacity:

$$\sum_{i=1}^N outParts[p][i] \leq capacity \quad \forall p : 1 \leq p \leq M$$

- Each qubit must be assigned to only one partition:

$$\sum_{p=1}^M outParts[p][i] = 1 \quad \forall i : 1 \leq i \leq N$$

- If two qubits  $q_i$  and  $q_j$  are assigned to different partitions,  $\text{cut}[i][j]$  must be set to 1:

$$\begin{aligned} \text{outParts}[p][i] - \text{outParts}[p][j] &\leq \text{cut}[i][j] \\ \forall i, j : 1 \leq i, j \leq N \text{ and } \forall p : 1 \leq p \leq M \end{aligned}$$

Suppose that qubits  $q_i$  and  $q_j$  are assigned to different partitions  $p_i$  and  $p_j$ , respectively. For  $p = p_i$  the left hand side of the above inequality is equal to 1 which forces  $\text{cut}[i][j]$  to be set to one. On the other hand, when  $q_i$  and  $q_j$  are in the same partitions, the left hand side of the inequality is zero for all partitions. In this case, the cost function forces  $\text{cut}[i][j]$  to be zero to minimize the cost.

- If the partitioning algorithm changes the partition of a qubit  $q_i$ ,  $\text{migrate}[i]$  must be set to 1:

$$\begin{aligned} \text{outParts}[p][i] - \text{prevParts}[p][i] &\leq \text{migrate}[i] \\ \forall i : 1 \leq i \leq N \text{ and } \forall p : 1 \leq p \leq M \end{aligned}$$

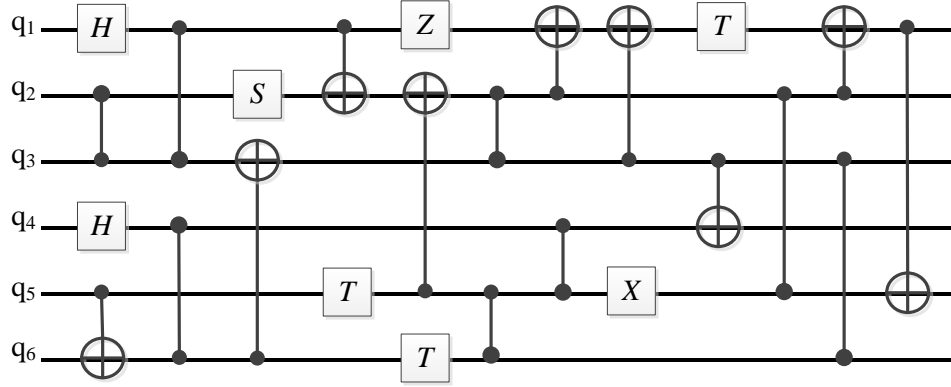
Suppose that the partition of qubit  $q_i$  has been changed from  $p_{\text{prev}}$  and  $p_i$ . Therefore, for  $p = p_i$  the left hand side of the above inequality is equal to 1 which forces  $\text{migrate}[i]$  to be set to 1. On the other hand, when  $q_i$  stays in its previous partition, the left hand side of the inequality is zero for all partitions. In this case, the cost function forces  $\text{migrate}[i]$  to be zero to minimize the cost.

#### 4.2. An example

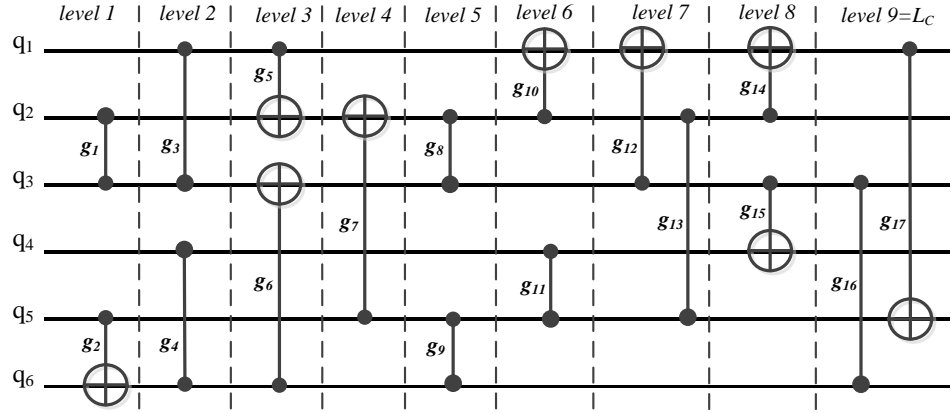
In this section, the proposed approach is explained by an example. Figure 4 shows a quantum circuit consisting of 6 qubits and 25 gates. Our algorithm partitions the circuit into two parts each containing three qubits. Let the window width be  $L_W = 3$ , window weight be  $W = \{w_1, w_2, w_3\} = \{3, 2, 1\}$ , and  $w_p = 2$ . The algorithm removes single-qubit gates from the circuit and then levelizes it in the first step, as shown in Figure 5. During the next step, the window is laid on the circuit starting from the first level. Figure 6 and Figure 7 show all steps of the algorithm. In the second column, the sub-circuit  $C_L$  is depicted. The corresponding interaction graph of  $C_L$  and  $P_L$ , i.e., the output of  $\text{subPartitioning}(C_L)$ , are shown in the third column. The last column contains the gates of level  $L$  ( $G_L$ ), the qubits that should be teleported using teledata and the gates that should be applied remotely by telegate, respectively.

For the graph of  $C_1$ , the gate  $g_1$  in the first level generates an edge between vertices  $q_2$  and  $q_3$  with the weight of 3 ( $w_1$ ). Similarly, there are edges  $E(q_4, q_6) = 2$  and  $E(q_3, q_6) = 1$  corresponding to gates  $g_4$  and  $g_6$  at levels 2 and 3, respectively. It should be noted that there is no edge between qubit and partition vertices in the graph of  $C_1$  because no qubit has been assigned to any partition yet.  $\text{subPartitioning}(C_1)$  partitions the qubits of  $C_1$  into two parts  $p_1 = \{q_1, q_2, q_3\}$  and  $p_2 = \{q_4, q_5, q_6\}$ , denoted by  $P_1 = \{p_1, p_1, p_1, p_2, p_2, p_2\}$  in Figure 6 and Figure 7.  $P_1$  is the initial partitioning of qubits and thus no teledata is required. In the next step, it is determined how the gates of  $G_1$ , i.e.,  $g_1$  and  $g_2$  should be applied. Both  $g_1$  and  $g_2$  can be applied locally because their qubits are assigned to the same partition. For the next levels, each qubit





**Figure 4.** An example circuit to partition into two parts



**Figure 5.** The leveled circuit after removing single-qubit gates

vertex is connected to the vertex corresponding to its previous partition with weight 2 ( $w_p$ ). In step 3, since  $q_3$  and  $q_6$  are assigned to different partitions,  $g_6$  must be applied remotely by telegate. In step 8,  $subPartitioning(C_8)$  partitions the qubits of the circuit into  $p_1 = \{q_1, q_2, q_5\}$  and  $p_2 = \{q_4, q_3, q_6\}$  which in comparison with the previous partitioning  $P_7$ ,  $q_3$  and  $q_5$  are assigned to different partitions. Therefore, these qubits will be transported using two teledata operations.

Our hybrid approach needs 5 communication operations including 2 teledata operations and 3 telegate ones while the best solution achieved by qubit partitioning approach or gate partitioning approach requires 6 communication operations. This example shows the superiority of our hybrid approach over both qubit partitioning and gate partitioning approaches.

## 5. Experimental Results

Our approach (WQCP) was implemented in C++ and CPLEX [49] was used as the ILP solver. It was run on a Core i7 CPU operating at 2.4 GHz with 8 GB of memory.

Single qubit gates and a two-qubit Clifford gate such as CNOT or CZ make a universal gate set for quantum computation. Therefore, the set CNOT, CZ and single

$L$	sub-circuit $C_L$	graph of $C_L$	$G_L$
		output of $\text{subPartitioning}(C_L)$	teledata telegate
1			$g_1$ $g_2$
		$P_1 = \{p_1, p_1, p_1, p_2, p_2, p_2\}$	-
2			$g_3$ $g_4$
		$P_2 = \{p_1, p_1, p_1, p_2, p_2, p_2\}$	-
3			$g_5$ $g_6$
		$P_3 = \{p_1, p_1, p_1, p_2, p_2, p_2\}$	$g_6$
4			$g_7$
		$P_4 = \{p_1, p_1, p_1, p_2, p_2, p_2\}$	$g_7$
5			$g_8$ $g_9$
		$P_5 = \{p_1, p_1, p_1, p_2, p_2, p_2\}$	-

Figure 6. Steps 1 to 5 of our approach to partition the example circuit of 4

$L$	sub-circuit $C_L$	graph of $C_L$	$G_L$
		output of $\text{subPartitioning}(C_L)$	teledata telegate
6			$g_{10}$ $g_{11}$
		$P_6 = \{p_1, p_1, p_1, p_2, p_2, p_2\}$	-
7			$g_{12}$ $g_{13}$
		$P_7 = \{p_1, p_1, p_1, p_2, p_2, p_2\}$	-
			$g_{13}$
8			$g_{14}$ $g_{15}$
		$P_8 = \{p_1, p_1, p_2, p_2, p_1, p_2\}$	-
			$q_3$ $q_5$
9			$g_{16}$ $g_{17}$
		$P_9 = \{p_1, p_1, p_2, p_2, p_1, p_2\}$	-
			-

Figure 7. Steps 6 to 9 of our approach to partition the example circuit of Figure 4

qubit gates was chosen as the gate library. To evaluate the performance of WQCP, it was applied to some benchmark circuits from [50] (the first nine circuits in the tables), Revlib [51] (the circuits from 10 to 15), some quantum error-correction encoding circuits [52] (the circuits from 16 to 25), and  $n$ -qubit quantum Fourier transform circuits ( $QFT$ ) [53] where  $n \in \{16, 32, 64, 128, 256\}$ .

These circuits may include some gates out of the gate library that are synthesized into the gates of the library based on the method proposed in [54].

The window length  $L_W$  may potentially have a high effect on the result. Table 2 compares the communication counts obtained by WQCP for different window lengths

where window weight  $W = \{L_W, L_W - 1, \dots, 1\}$  and  $w_p = L_W - 1$ . The number of qubits, the multi-qubit depth of each benchmark and the number of partitions are shown in the second column. The third column contains the type of teleportation, which can be teledata or telegate. The best obtained results are marked in bold. It is worth noting that the window weight  $W$  and  $w_p$  were chosen experimentally and different results may be achieved by changing them.

Table 2: Experimental results (number of teleportations) obtained by WQCP for the benchmark circuits and different window lengths  $L_W$

#	Benchmark	# qubits Depth	TP type	$L_W$						
		# parts		5	6	7	8	9	12	15
1	2of5-D1	6	Telegate	11	16	31	31	30	30	30
		98	Teledata	38	30	10	10	10	10	10
		2	Total	49	46	41	41	<b>40</b>	<b>40</b>	<b>40</b>
2	2-4dec	6	Telegate	9	9	9	9	9	8	9
		19	Teledata	6	2	2	2	4	4	4
		3	Total	15	<b>11</b>	<b>11</b>	<b>11</b>	13	12	13
3	6sym	10	Telegate	5	4	5	4	6	9	8
		42	Teledata	18	16	14	12	12	10	8
		2	Total	23	20	19	<b>16</b>	18	19	<b>16</b>
4	9sym	12	Telegate	5	17	20	24	29	20	24
		71	Teledata	31	20	20	16	16	16	17
		3	Total	<b>36</b>	37	40	40	45	<b>36</b>	41
5	Ham15-D3	15	Telegate	43	40	56	51	59	56	57
		177	Teledata	70	63	49	54	43	48	44
		4	Total	113	103	105	105	102	104	<b>101</b>
6	Cycle17-3	20	Telegate	246	255	291	589	83	702	740
		8561	Teledata	2261	2217	2118	1722	1657	1380	1288
		3	Total	2507	2472	2409	2311	2340	2082	<b>2028</b>
7	8bitadder	24	Telegate	11	21	19	26	22	46	41
		106	Teledata	87	73	65	58	69	42	48
		6	Total	98	94	<b>84</b>	<b>84</b>	91	88	89
8	Hwb50	56	Telegate	325	386	317	287	325	293	400
		4994	Teledata	1319	1380	1162	1232	1057	1008	936
		5	Total	1644	1766	1479	1519	1382	<b>1301</b>	1336
9	Hwb100	107	Telegate	1075	676	982	1122	597	753	673
		15923	Teledata	3446	3151	3117	2805	2295	2029	1840
		7	Total	4521	3827	4099	3927	2892	2782	<b>2513</b>
10	rd32_272	5	Telegate	2	1	1	1	1	1	1
		5	Teledata	4	6	6	6	6	6	6
		2	Total	<b>6</b>	7	7	7	7	7	7
11	ham7_106	7	Telegate	26	26	26	26	26	26	27
		38	Teledata	4	4	4	4	4	7	6
		4	Total	<b>30</b>	<b>30</b>	<b>30</b>	<b>30</b>	<b>30</b>	33	33
12	rd53_139	8	Telegate	6	3	4	2	3	6	9
		8	Teledata	8	10	10	12	10	6	4
		2	Total	14	13	14	14	13	<b>12</b>	13
13	rd53_311	13	Telegate	5	5	2	3	3	4	2
		19	Teledata	19	22	23	21	20	18	20
		3	Total	24	27	25	24	23	<b>22</b>	<b>22</b>
14	parity_247	17	Telegate	1	1	3	2	4	3	4
		16	Teledata	3	4	3	4	3	4	3
		3	Total	<b>4</b>	5	6	6	7	7	7
15	adder16_174	49	Telegate	2	2	2	2	2	3	3
		19	Teledata	9	7	10	11	11	8	7
		3	Total	11	<b>9</b>	12	13	13	11	10
16	[[10,3,3]]	10	Telegate	4	5	5	5	6	7	6
		25	Teledata	10	8	8	8	8	8	8
		2	Total	14	<b>13</b>	<b>13</b>	<b>13</b>	14	15	14
17	[[16,3,5]]	16	Telegate	17	16	18	27	28	30	34
		43	Teledata	34	38	33	21	21	19	17
		4	Total	51	54	51	<b>48</b>	49	49	51
18	[[21,1,7]]	21	Telegate	5	9	14	20	26	32	23
		58	Teledata	51	51	46	42	32	28	37

Table 2: Experimental results (number of teleportations) obtained by WQCP for the benchmark circuits and different window lengths  $L_W$  (continue)

#	Benchmark	# qubits	TP type	$L_W$						
		Depth # parts		5	6	7	8	9	12	15
19	[[24,3,7]]	3	Total	<b>56</b>	60	60	62	58	60	60
		24	Telegate	13	17	34	35	38	37	50
		84	Teledata	88	88	69	71	66	66	57
		4	Total	<b>101</b>	105	103	106	104	103	107
20	[[25,1,9]]	25	Telegate	19	22	21	21	22	26	43
		83	Teledata	79	70	68	68	66	67	53
		5	Total	98	92	89	89	<b>88</b>	93	96
21	[[27,1,9]]	27	Telegate	17	12	31	44	35	46	45
		110	Teledata	89	86	77	67	76	68	69
		4	Total	106	<b>98</b>	108	111	111	114	114
22	[[31,11,6]]	31	Telegate	18	28	42	39	45	60	61
		149	Teledata	127	129	110	99	105	101	94
		4	Total	145	157	152	<b>138</b>	150	161	155
23	[[33,1,9]]	33	Telegate	9	19	32	36	52	56	55
		153	Teledata	132	119	122	124	102	103	104
		5	Total	141	<b>138</b>	154	160	154	159	159
24	[[35,1,10]]	35	Telegate	10	22	21	31	43	48	62
		126	Teledata	114	111	99	100	100	102	105
		4	Total	124	133	<b>120</b>	131	143	150	167
25	[[40,3,10]]	40	Telegate	23	34	40	50	53	59	67
		172	Teledata	166	150	153	136	149	138	122
		4	Total	189	<b>184</b>	193	186	202	197	189
26	QFT16	16	Telegate	0	0	1	14	15	18	18
		56	Teledata	23	23	23	25	26	22	28
		3	Total	<b>23</b>	<b>23</b>	24	39	41	40	46
27	QFT32	32	Telegate	0	0	0	0	0	0	0
		120	Teledata	48	48	48	48	48	48	48
		4	Total	<b>48</b>	<b>48</b>	<b>48</b>	<b>48</b>	<b>48</b>	<b>48</b>	<b>48</b>
28	QFT64	64	Telegate	0	0	0	0	0	0	0
		248	Teledata	106	106	106	106	106	106	106
		6	Total	<b>106</b>	<b>106</b>	<b>106</b>	<b>106</b>	<b>106</b>	<b>106</b>	<b>106</b>
29	QFT128	128	Telegate	0	0	0	0	0	0	0
		504	Teledata	224	224	224	224	224	224	224
		8	Total	<b>224</b>	<b>224</b>	<b>224</b>	<b>224</b>	<b>224</b>	<b>224</b>	<b>224</b>
30	QFT256	256	Telegate	0	0	NA	NA	NA	NA	NA
		1016	Teledata	468	468	NA	NA	NA	NA	NA
		12	Total	<b>468</b>	<b>468</b>	NA	NA	NA	NA	NA

Table 3 compares the best results obtained by WQCP with the qubit and gate partitioning approaches. Two different algorithms based on qubit partitioning approach are considered. In the first one, that is denoted by QPILP, the qubit partitioning is modeled using ILP and solved by CPLEX solver. Although this method produces the optimal results of qubit partitioning, it is not scalable. The method proposed by Ahsan et al. [7, 32], denoted by QPGTA, is considered as the second algorithm for comparison. To implement gate partitioning approach (*GP*), WQCP was used where the window weight  $w_1$  was set to a very large number compared to the other weights, i.e.  $W \setminus \{w_1\}$  and  $w_p$ . By this, WQCP is forced to apply each multi-qubit locally without any telegate operation. Table 3 shows that WQCP decreases the communication cost, on average, by 37.6% and 21.4% in comparison to the qubit and gate partitioning approach, respectively. For the QFT circuits, the best partitioning approach is gate partitioning because of their particular structures and WQCP produces the same results as GP for these circuits.

Table 3: The partitioning results achieved by WQCP for the benchmark circuits compared with the gate and qubit partitioning approaches

#	Benchmark	TP type	GP	QPILP	QPGTA	WQCP	Improvement (%)		
							GP	QPILP	QPGTA
1	2of5-D1	Telegate	0	50	50	30	35	20	20
		Teledata	62	0	0	10			
		Total	62	50	50	40			
2	2-4dec	Telegate	0	13	18	9	59	15	38
		Teledata	27	0	0	2			
		Total	27	13	18	11			
3	6sym	Telegate	0	22	22	4	42	27	27
		Teledata	28	0	0	12			
		Total	28	22	22	16			
4	9sym	Telegate	0	57	72	5	16	37	50
		Teledata	43	0	0	31			
		Total	43	57	72	36			
5	Ham15-D3	Telegate	0	126	156	57	35	20	35
		Teledata	155	0	0	44			
		Total	155	126	156	101			
6	Cycle17-3	Telegate	0	3979	4323	740	15	49	53
		Teledata	2372	0	0	1288			
		Total	2372	3979	4323	2028			
7	8bitadder	Telegate	0	NA	147	19	18	NA	43
		Teledata	103	NA	0	65			
		Total	103	NA	147	84			
8	Hwb50	Telegate	0	NA	3247	293	44	NA	60
		Teledata	2323	NA	0	1008			
		Total	2323	NA	3247	1301			
9	Hwb100	Telegate	0	NA	4995	673	34	NA	50
		Teledata	3825	NA	0	1840			
		Total	3825	NA	4995	2513			
10	rd32_272	Telegate	0	10	10	2	25	40	40
		Teledata	8	0	0	4			
		Total	8	10	10	6			
11	ham7_106	Telegate	0	32	32	26	57	6	6
		Teledata	71	0	0	4			
		Total	71	32	32	30			
12	rd53_139	Telegate	0	17	17	6	25	29	29
		Teledata	16	0	0	6			
		Total	16	17	17	12			
13	rd53_311	Telegate	0	41	41	4	15	46	46
		Teledata	26	0	0	18			
		Total	26	41	41	22			
14	parity_247	Telegate	0	11	11	1	20	63	63
		Teledata	5	0	0	3			
		Total	5	11	11	4			
15	adder16_174	Telegate	0	17	17	2	30	47	47
		Teledata	13	0	0	7			
		Total	13	17	17	9			
16	[[10,3,3]]	Telegate	0	16	18	5	27	18	27
		Teledata	18	0	0	8			
		Total	18	16	18	13			
17	[[16,3,5]]	Telegate	0	NA	55	27	30	NA	12
		Teledata	69	NA	0	21			
		Total	69	NA	55	48			
18	[[21,1,7]]	Telegate	0	NA	80	5	14	NA	30
		Teledata	65	NA	0	51			
		Total	65	NA	80	56			
19	[[24,3,7]]	Telegate	0	NA	140	13	13	NA	28
		Teledata	117	NA	0	88			
		Total	117	NA	140	101			
20	[[25,1,9]]	Telegate	0	NA	111	22	24	NA	20
		Teledata	117	NA	0	66			
		Total	117	NA	111	88			
21	[[27,1,9]]	Telegate	0	NA	163	12	15	NA	40
		Teledata	116	NA	0	86			
		Total	116	NA	163	98			
22	[[31,11,6]]	Telegate	0	NA	231	39	20	NA	40

Table 3: The partitioning results achieved by WQCP for the benchmark circuits compared with the gate and qubit partitioning approaches (continue)

#	Benchmark	TP type	GP	QPILP	QPGTA	WQCP	Improvement (%)		
							GP	QPILP	QPGTA
23	[[33,1,9]]	Teledata	172	NA	0	99	12	NA	37
		Total	172	NA	231	138			
		Telegate	0	NA	220	19			
		Teledata	157	NA	0	119			
		Total	157	NA	220	138			
		Telegate	0	NA	267	21			
24	[[35,1,10]]	Teledata	132	NA	0	99	9	NA	55
		Total	132	NA	267	120			
		Telegate	0	NA	328	34			
		Teledata	196	NA	0	150			
25	[[40,3,10]]	Total	196	NA	328	184	6	NA	44
		Telegate	0	87	94	0			
		Teledata	23	0	0	23			
26	QFT16	Total	23	87	94	23	0	73	75
		Telegate	0	NA	165	0			
		Teledata	48	NA	0	48			
27	QFT32	Total	48	NA	165	48	0	NA	70
		Telegate	0	NA	275	0			
		Teledata	106	NA	0	106			
28	QFT64	Total	106	NA	275	106	0	NA	61
		Telegate	0	NA	385	0			
		Teledata	224	NA	0	224			
29	QFT128	Total	224	NA	385	224	0	NA	41
		Telegate	0	NA	605	0			
		Teledata	468	NA	0	468			
30	QFT256	Total	468	NA	605	468	0	NA	22
		Telegate	0	NA	605	0			

The runtime of our WQCP approach in comparison to the previous approaches (GP, QPILP, and QPGTA) is reported in Table 4. Since our approach and GP both use the same approach, their runtimes are approximately the same. QPILP is faster than ours for small circuits because the ILP qubit partitioning is run only once in QPILP while WQCP calls it for each level of a circuit. However, by increasing the size of the circuits, QPILP's resource overhead grows exponentially and it fails to generate an output as the memory runs out. Finally, QPGTA is the fastest approach. Its runtime is less than one second for all benchmark circuits.

The runtime of WQCP grows by increasing the number of qubits, the number of partitions, the window size, and the multi-qubit depth of a circuit. Since our approach calls the ILP solver only for each window, the runtime of our tool is manageable and, as it can be seen, it is applicable to large circuits such as Hwb100 and QFT256. Although the number of qubits in a window is the same as the total number of qubits in the circuit, the adjacency matrix is sparse for a sub-circuit contained in a window. This feature enables our approach to partition large circuits while other approaches using the ILP solver without windowing strategy fails to produce an output. Moreover, using some techniques such as hierarchical partitioning and utilizing a faster algorithm rather than ILP to implement  $\text{subPartitioning}(C_l)$  can accelerate WQCP probably at the cost of decreasing the quality of results.

**Table 4.** The runtimes of WQCP (in milisecond) for the benchmark circuits compared with the gate and qubit partitioning approaches

Benchmark	GP	QPILP	QPGTA	WQCP
<b>2of5-D1</b>	9122	305	3	7500
<b>2-4dec</b>	2259	122	4	1762
<b>6sym</b>	2956	158	4	3774
<b>9sym</b>	8379	605	6	7231
<b>Ham15-D3</b>	18631	140087	6	20710
<b>Cycle17-3</b>	717058	148226	12	871793
<b>8bitadder</b>	11867	1490	6	21051
<b>Hwb50</b>	758887	NA	20	865777
<b>Hwb100</b>	4235861	NA	39	4266630
<b>rd32_272</b>	1445	140	4	1392
<b>ham7_106</b>	5788	215	5	6580
<b>rd53_139</b>	2573	218	6	2266
<b>rd53_311</b>	6600	520	5	7343
<b>parity_247</b>	1510	166	5	1938
<b>adder16_174</b>	9978	2095	15	9724
<b>[[10,3,3]]</b>	1959	334	7	1835
<b>[[16,3,5]]</b>	9188	NA	5	10802
<b>[[21,1,7]]</b>	7491	NA	6	7294
<b>[[24,3,7]]</b>	18235	NA	7	14239
<b>[[25,1,9]]</b>	17441	NA	14	22624
<b>[[27,1,9]]</b>	20741	NA	14	26925
<b>[[31,11,6]]</b>	26014	NA	11	38533
<b>[[33,1,9]]</b>	27008	NA	10	27712
<b>[[35,1,10]]</b>	28471	NA	12	24308
<b>[[40,3,10]]</b>	41794	NA	14	34374
<b>QFT16</b>	11352	179462	12	5268
<b>QFT32</b>	16216	NA	18	15491
<b>QFT64</b>	34099	NA	37	37288
<b>QFT128</b>	131541	NA	57	150314
<b>QFT256</b>	1.386e6	NA	156	1.391e6

## 6. Conclusion

The main challenge of distributed quantum computing is costly communications between processing units which may be an order of magnitude more time consuming and error prone than logical operations. In this paper, we proposed an automated window-based partitioning method called WQCP to minimize such communications. The proposed method reduces the communication cost by about 29.5% in comparison with the best approaches reported in the literature.

Although the execution time of WQCP is more than existing approaches, it is not a challenge as it runs offline before actual computation. Moreover, one may speed up WQCP by utilizing a faster algorithm rather than the ILP one. Furthermore, in this paper the window weights and  $w_p$  are fixed and set manually based on our experiments. While these weights may highly effect the results. Automating window weight setting based on the input circuit can be followed as future work.

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