Applying and Expanding the VOQC Toolkit

KESHA HIETALA, University of Maryland, USA
LIYI LI, University of Maryland, USA
AKSHAJ GAUR, Poolesville High School, USA
AARON GREEN, University of Maryland, USA
ROBERT RAND, University of Chicago, USA
XIAODI WU, University of Maryland, USA
MICHAEL HICKS, University of Maryland, USA

This abstract presents recent extensions to voqc, a verified optimizer for quantum circuits, first presented at POPL 2021 [Hietala et al. 2021b]. All code described in this abstract is freely available online.¹

1 OVERVIEW

voqc [Hietala et al. 2021b] (pronounced “vox”) is a compiler for quantum circuits, in the style of tools like Qiskit [Aleksandrowicz et al. 2019], tket [Cambridge Quantum Computing Ltd 2019], Quilc [Rigetti Computing 2019], and Cirq [Developers 2021]. What makes voqc different from these tools is that it has been formally verified in the Coq proof assistant [Coq Development Team 2019].

voqc source programs are expressed in sqir, a simple quantum intermediate representation, which has a precise mathematical semantics. We use Gallina, Coq’s programming language, to implement voqc transformations over sqir programs, and use Coq to prove the source program’s semantics are preserved. We then extract these Gallina definitions to OCaml, and compile the OCaml code to a library that can operate on standard-formatted circuits.

voqc, and sqir, were built to be general-purpose. For example, while we originally designed sqir for use in verified optimizations, we subsequently found sqir could also be suitable for writing, and proving correct, source programs [Hietala et al. 2021a]. We have continued to develop the voqc codebase to expand its reach and utility.

In this abstract, we present new extensions to voqc as an illustration of its flexibility. These include support for calling voqc transformations from Python, added support for new gate sets and optimizations, and the extension of our notion of correctness to include mapping-preservation, which allows us to apply optimizations after mapping, reducing the cost introduced by making a program conform to hardware constraints.

2 PYVOQC

In order to make voqc compatible with existing Python-based frameworks for compiling quantum programs (e.g. Qiskit, pytket, Quilc, Cirq), we provide a Python wrapper (dubbed pyvoqc) around the voqc OCaml library. To interface between Python and OCaml, we wrap the OCaml code in a C library (following standard conventions [INRIA 2021]) and call to this C library using Python’s

¹Software links:
- Our Coq definitions and proofs are available at https://github.com/inQWIRE/SQIR.
- Our OCaml library is available at https://github.com/inQWIRE/mlvoqc and can be installed with “opam install voqc”.
- Documentation on the OCaml library interface is available at https://inqwire.github.io/mlvoqc/voqc/Voqc/index.html.
- Our Python bindings and tutorials are available at https://github.com/inQWIRE/pyvoqc.

Authors’ addresses: Kesha Hietala, University of Maryland, USA, kesh@cs.umd.edu; Liyi Li, University of Maryland, USA, liyili2@umd.edu; Akshaj Gaur, Poolesville High School, USA, akshajgaur@gmail.com; Aaron Green, University of Maryland, USA, agreen19@umd.edu; Robert Rand, University of Chicago, USA, rand@uchicago.edu; Xiaodi Wu, University of Maryland, USA, xwu@cs.umd.edu; Michael Hicks, University of Maryland, USA, mwh@cs.umd.edu.
Table 1. Gate sets used in voqc. \( r \) is a real parameter and \( q \) is a rational parameter.

<table>
<thead>
<tr>
<th>Gate Set</th>
<th>Single-qubit gates</th>
<th>Two-qubit gates</th>
<th>Three-qubit gates</th>
</tr>
</thead>
<tbody>
<tr>
<td>Standard</td>
<td>I, X, Y, Z, H, S, T, Sdg, Tdg, Rx(( r )), Ry(( r )), Rz(( r )), Rzq(( q )), U1(( r )), U2(( r, r )), U3(( r, r, r ))</td>
<td>CX, CZ, SWAP</td>
<td>CCX, CCZ</td>
</tr>
<tr>
<td>RzQ</td>
<td>X, H, Rzq(( q ))</td>
<td></td>
<td></td>
</tr>
<tr>
<td>IBM</td>
<td>U1(( r )), U2(( r, r )), U3(( r, r, r ))</td>
<td>CX</td>
<td></td>
</tr>
</tbody>
</table>


c types [Python Software Foundation 2021]. For convenience, we have written Python code that makes voqc look like an optimization pass in IBM’s Qiskit or Google’s Cirq, allowing us to take advantage of these frameworks’ utilities for quantum programming (e.g. constructing and printing circuits, unverified optimizations and mapping routines). We show an example of using voqc as a Qiskit pass in Section 3.

3 SUPPORT FOR ADDITIONAL GATE SETS

When we first presented voqc we defined our optimizations over the “RzQ” gate set [Hietala et al. 2021b, §4.2], but argued that our work could be applied to other gate sets as well since many of our utility definitions and proofs are gate-set independent. Here, we demonstrate this extensibility for two new gate sets: the “standard” gate set and the “IBM” gate set, both shown in Table 1. The standard gate set is used for parsing and aims for completeness: Instead of having to translate a \( T \) gate in the source program to the semantically equivalent Rz(\( \pi/4 \)), we can translate it directly to T.

Likewise, we can translate the three-qubit CCX gate directly to CCX, rather than decomposing it into a series of one- and two-qubit gates (potentially incorrectly).

The IBM gate set is the default basis for the Qiskit compiler.\(^2\) It includes the two-qubit controlled-NOT (CX) gate, along with three parameterized single-qubit gates:

\[
U_1(\lambda) = \begin{pmatrix} 1 & 0 \\ 0 & e^{i\lambda} \end{pmatrix}, \quad U_2(\phi, \lambda) = \frac{1}{\sqrt{2}} \begin{pmatrix} e^{i\phi} & -e^{i\lambda} \\ e^{i(\phi+\lambda)} & e^{i\lambda} \end{pmatrix}, \quad U_3(\theta, \phi, \lambda) = \begin{pmatrix} \cos(\theta/2) & -e^{i\lambda} \sin(\theta/2) \\ e^{i\phi} \sin(\theta/2) & e^{i(\phi+\lambda)} \cos(\theta/2) \end{pmatrix}.
\]

One interesting property of this gate set (which does not hold of our original RzQ gate set) is that any sequence of single-qubit gates can be combined into a single gate. This allows us to implement an optimization (which Qiskit calls Optimize\( 1q\)Gates) that merges all adjacent single-qubit gate by applying rules like \( U_1(\lambda_1); U_1(\lambda_2) \rightarrow U_1(\lambda_1 + \lambda_2) \) and \( U_1(\lambda_1); U_2(\phi, \lambda_2) \rightarrow U_2(\lambda_2, \lambda_1 + \phi) \).

The most complicated rule for merging gates is the one for combining a \( U_2 \) and \( U_3 \) gate or two \( U_3 \) gates. In this case, the two gates are first converted into a sequence of Euler rotations [Euler 1776] about the \( y \)- and \( z \)-axes, e.g. \( U_3(\theta, \phi, \lambda) = R_z(\phi) \cdot R_y(\theta) \cdot R_z(\lambda) \). We will call this a YZ rotation.

Next, local identities are applied to combine the two YZ rotations into a single YZYZ rotation. Then the interior YZY rotation is converted to a new YZ rotation, yielding a ZZYZZ rotation. Finally, this is simplified to a YZ rotation, which can be represented as a \( U_3 \) gate. For example,

\(^2\)Actually, the \( U_1, U_2, \) and \( U_3 \) gates are used in many quantum compilers. We called this the “IBM” gate set because, at the time, we were aiming to verify a Qiskit optimization.
from qiskit import QuantumCircuit
from pyvoqc.qiskit.voqc_pass import QiskitVOQC
from qiskit.transpiler import PassManager

# create a circuit using Qiskit's interface
circ = QuantumCircuit(2)
circ.x(0)
circ.t(0)
circ.cx(0, 1)
circ.t(0)
circ.tdg(1)
print("Before Optimization:")
print(circ)

# create a Qiskit PassManager
pm = PassManager()

# decompose CZ gate
pm.append(QiskitVOQC(“decompose_to_cnot”))
new_circ = pm.run(circ)
print("After 'decompose_to_cnot':")
print(new_circ)

# run optimizations from Nam et al.
pm.append(QiskitVOQC(“optimize_nam”, “replace_rzq”))
new_circ = pm.run(circ)
print("After 'optimize_nam':")
print(new_circ)

# run IBM gate merging
pm.append(QiskitVOQC(“optimize_ibm”))
new_circ = pm.run(circ)
print("After 'optimize_ibm':")
print(new_circ)

Fig. 1. Example of using voqc as a Qiskit pass. The output from the script on the left is shown on the right.

here is the process for combining two $U_3$ gates:

$$U_3(\theta_1, \phi_1, \lambda_1); U_3(\theta_2, \phi_2, \lambda_2) = R_z(\phi_2) \cdot R_y(\theta_2) \cdot R_z(\lambda_2) \cdot R_z(\phi_1) \cdot R_y(\theta_1) \cdot R_z(\lambda_1) = R_z(\phi_2) \cdot \left[ R_y(\theta_2) \cdot R_z(\lambda_2 + \phi_1) \cdot R_y(\theta_1) \right] \cdot R_z(\lambda_1) = R_z(\phi_2 + \gamma) \cdot R_y(\beta) \cdot R_z(\alpha + \lambda_1) = U_3(\beta, \phi_2 + \gamma, \alpha + \lambda_1)$$

where $\alpha, \beta, \gamma$ satisfy $R_y(\theta_2) \cdot R_z(\lambda_2 + \phi_1) \cdot R_y(\theta_1) = R_z(\gamma) \cdot R_y(\beta) \cdot R_z(\alpha)$.

Importantly, we were able to define and verify this optimization using the same infrastructure we had developed for our original RzQ gate set, showing that our framework is indeed extensible. The most difficult part of the proof was showing the correctness of the gate combination rules. In particular, the method for converting from a YZY to YZ$\bar{Y}$ rotation (shown in Figure 2) was challenging to verify because it contains many cases, all of which involve complicated trigonometric expressions. To our knowledge, we are the first to formally verify this method in a proof assistant.

In total, VOQC supports the 9 optimizations listed in Table 2. We show the effect of applying optimize_nam and optimize_ibm (as Qiskit passes) on an example circuit in Figure 1. In addition, VOQC provides a simple_map function that takes as input a circuit, a description of the underlying architecture connectivity, and an initial mapping from the circuit’s qubits to machine qubits, and returns a program that respects the constraints of the architecture (see Section 4 for more details). Our simple_map routine is effectively the same as Qiskit’s BasicSwap pass [Qiskit Development Team 2021].
\[
\text{Definition } \text{rm}02 (x \ y \ z : R) := \sin x \cdot \cos z + \cos x \cdot \cos y \cdot \sin z.
\]
\[
\text{Definition } \text{rm}12 (x \ y \ z : R) := \sin y \cdot \sin z.
\]
\[
\text{Definition } \text{rm}22 (x \ y \ z : R) := \cos x \cdot \cos z - \sin x \cdot \cos y \cdot \sin z.
\]
\[
\text{Definition } \text{rm}10 (x \ y \ z : R) := \sin y \cdot \cos z.
\]
\[
\text{Definition } \text{rm}11 (x \ y \ z : R) := \cos y.
\]
\[
\text{Definition } \text{rm}20_{-\text{minus}} (x \ y \ z : R) := \cos x \cdot \sin z + \sin x \cdot \cos y \cdot \cos z.
\]
\[
\text{Definition } \text{rm}21 (x \ y \ z : R) := \sin x \cdot \sin y.
\]
\[
\text{Definition } \text{atan2} (y \ x : R) := \begin{cases} 
\text{if } 0 < y \text{ then } \text{atan}(y/x) \\
\text{else if } x < 0 \text{ then if negb}(y < 0) \text{ then } \text{atan}(y/x) + \pi \text{ else if } y < 0 \text{ then } -\pi/2 \text{ else } 0.
\end{cases}
\]
\[
\text{Definition } \text{yzy_to_zyz} (x \ y \ z : R \ast R \ast R) := \begin{cases} 
\text{if } \text{rm}22 x y z < 1 \text{ then if } -1 < \text{rm}22 x y z \\
\text{then } (\text{atan2}(\text{rm}12 x y z)(\text{rm}02 x y z), \\
\text{acos}(\text{rm}22 x y z), \\
\text{atan2}(\text{rm}21 x y z)(\text{rm}20_{-\text{minus}} x y z)) \\
\text{else } (- \text{atan2}(\text{rm}10 x y z)(\text{rm}11 x y z), \pi, 0) \\
\text{else } (\text{atan2}(\text{rm}10 x y z)(\text{rm}11 x y z), 0, 0).
\end{cases}
\]
\[
(* \text{Correctness property: } *)
\]
\[
\text{Lemma } \text{yzy_to_zyz_correct} : \forall \theta_1 \xi \theta_2 \xi_1 \theta \xi_2, \\
\text{yzy_to_zyz } \theta_1 \xi \theta_2 = (\xi_1, \theta, \xi_2) \rightarrow \\
\text{y_rotation } \theta_2 \times \text{phase_shift } \xi \times \text{y_rotation } \theta_1 \\
\propto \text{phase_shift } \xi_2 \times \text{y_rotation } \theta \times \text{phase_shift } \xi_1.
\]

Fig. 2. Code for converting a YZY rotation to a ZYZ rotation.

4 INTERLEAVING MAPPING AND OPTIMIZATION

Near-term machines only allow two-qubit gates to be applied between certain pairs of qubits and in particular orientations. For example, in IBM’s 5-qubit Tenerife machine (shown on the right), a CX gate may be applied with Q4 as the control and Q2 as the target, but not the reverse. No two-qubit gate is possible between physical qubits Q4 and Q1. So the program CX Q4 Q1 will need to be transformed to, e.g., SWAP Q2 Q4; CX Q2 Q1 in order to be executed on the machine.\(^3\)

\(^3\)The SWAP gate will be decomposed to CX Q4 Q2; CX Q2 Q4; CX Q4 Q2, which is transformed into CX Q4 Q2; H Q2; H Q4; CX Q4 Q2; H Q2; H Q4; CX Q4 Q2 to respect architecture constraints.

\(^3\)Circuit mapping automates this process, taking as input a circuit and architecture connectivity graph, and returning a transformed circuit that respects the constraints of the architecture [Saeedi et al. 2011; Zulehner et al. 2017]. Circuit mapping increases the number of gates, typically adding many CX and H gates to perform SWAPs between qubits.

Fig. 3. Connectivity on IBM’s 5-qubit Tenerife machine.
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Table 2. Optimizations available in voqc.

<table>
<thead>
<tr>
<th>Name</th>
<th>Description</th>
<th>Gate Set</th>
</tr>
</thead>
<tbody>
<tr>
<td>not_propagation</td>
<td>[Hietala et al. 2021b, §4.3]</td>
<td>RzQ</td>
</tr>
<tr>
<td>hadamard_reduction</td>
<td>[Hietala et al. 2021b, §4.4]</td>
<td>RzQ</td>
</tr>
<tr>
<td>cancel_single_qubit_gates</td>
<td>[Hietala et al. 2021b, §4.3]</td>
<td>RzQ</td>
</tr>
<tr>
<td>cancel_two_qubit_gates</td>
<td>[Hietala et al. 2021b, §4.3]</td>
<td>RzQ</td>
</tr>
<tr>
<td>merge_rotations</td>
<td>[Hietala et al. 2021b, §4.4]</td>
<td>RzQ</td>
</tr>
<tr>
<td>optimize_nam</td>
<td>Applies all RzQ optimizations in the ordering described in Hietala et al. [2021b, §4.6]</td>
<td>RzQ</td>
</tr>
<tr>
<td>optimize_1q_gates</td>
<td>Implementation of Qiskit’s Optimize1qGates [Qiskit Development Team 2021]</td>
<td>IBM</td>
</tr>
<tr>
<td>cx_cancellation</td>
<td>Implementation of Qiskit’s CXCancellation [Qiskit Development Team 2021]</td>
<td>IBM</td>
</tr>
<tr>
<td>optimize_ibm</td>
<td>Applies optimize_1q_gates followed by cx_cancellation</td>
<td>IBM</td>
</tr>
</tbody>
</table>

It is desirable to reduce this overhead by applying optimization after mapping. However, this is only worthwhile if the optimization preserves the guarantee from mapping that all CX gates are allowed by the connectivity graph. We have verified that all of the optimizations in Table 2, except hadamard_reduction, preserve connectivity guarantees. We call this property *mapping-preservation*, in contrast to the standard property that we prove, *semantics-preservation*, which says that an optimization does not change the behavior (“semantics”) of the input program.

5 ONGOING WORK

We are working to extend voqc with more gates, optimizations, and mapping routines, taking inspiration from frameworks like Qiskit, tket, Quilc, and Cirq. We are especially interested in implementing and verifying more sophisticated circuit mappers and adding support for *approximate* optimizations that do not preserve semantics exactly, but instead return a lower-cost circuit with similar behavior (e.g. [Peterson 2021]). Our current mapping routine is quite simple compared to state-of-the-art mappers, which involve complex subroutines like A* search [Zulehner et al. 2017] or Steiner tree approximations [Nash et al. 2020]. To avoid verifying the entirety of these algorithms, we are exploring approaches to verified translation validation of their components.

One major limitation of our current work with the IBM gate set is that we have proved optimizations correct for gates that use Coq real numbers as parameters. Because Coq reals are axiomatized, there is no way to extract our Coq definitions to OCaml without providing an implementation of real arithmetic. For simplicity (and compatibility with existing frameworks), we have chosen to extract Coq reals to OCaml floats. This is not ideal because it allows for the possibility of rounding error not accounted for in our proofs. We previously avoided this issue by using rational gate parameters (which can be extracted to OCaml multi-precision rationals), but this is not sufficient for the IBM single-qubit gate optimization, which involves trigonometric functions that are not defined over rationals (Figure 2). One possible solution is to verify our optimizations over gates that use Coq float parameters (e.g. using the Floqc library [Boldo and Melquiond 2011]).

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