NL HUNGARY HUNGARY HUNGARY Synthesized via data-flow engines





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Laboratory of Quantum Computer Simulators





Zoltán Zimborás • many-body phisics • quantum computing

• quantum information



Zoltán Kolarovszki

- software engeneering
- optical quantum computing
- Python programming





Peter Rakyta • condensed matters

 parallel, hardware oriented programming

grog



Gregory Morse • software engineer

- machine learning
- parallel, hardware oriented programming



Gábor Vattay

- Complex Systems
- Quantum chaos





Tamás Kozsikfunctional programmingprogramming languages



Ágoston Kaposi

- algebraic and differential topology
- mathematical network modelling
- C++, python programming

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Quantum gate decompositionQuantum Information
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HUNGARYquantum program (unitary): $UU^{\dagger} = 1$

preserves the norm of the state: $\langle U\Psi|U\Psi\rangle = \langle \Psi|U^{\dagger}U|\Psi\rangle = \langle \Psi|\Psi\rangle$

 $\hat{U} = \hat{U}_{11} \cdot \hat{U}_{10} \cdot \hat{U}_9 \cdot \hat{U}_8 \cdot \hat{U}_7 \cdot \hat{U}_6 \cdot \hat{U}_5 \cdot \hat{U}_4 \cdot \hat{U}_3 \cdot \hat{U}_2 \cdot \hat{U}_1$



Optimized quantum circuit synthesis



How to find an optimal gate decomposition?

- fewest gate count?
- smallest depth?

Available gate decomposition utilities:

- Quantum Fast Approximate Synthesis Tool (QFAST)
- QSearch + LEAP

(Lawrence Berkeley National Laboratory)



• UniversalQCompiler (incorporated into QISKIT)

(ETH Zürich, University of York, TUM)

CQ T|ket>: A Retargetable Compiler for NISQ Devices (Cambridge Quantum Computing Ltd., University of Strathclyde

How close is an approximation to the exact one?

exact evolution: U $|\psi(U)\rangle \coloneqq U|\psi\rangle$

approximate evolution: V

$$|\psi(V)\rangle \coloneqq V|\psi\rangle$$

The fidelity of the approximation: $\overline{F}(U,V) \coloneqq \int_{\psi} |\langle \psi(V) | \psi(U) \rangle|^2 \, \mathrm{d}\psi$

average taken over the Haar distribution

The cost function of the optimization:

Hilbert-Schmidt test:

$$C_{HST}(U,V) = 1 - \frac{1}{d^2} \left| \operatorname{Tr} \left(V^{\dagger} U \right) \right|^2$$

$$\overline{F}(U,V) = 1 - \frac{d}{d+1} C_{HST}(U,V)$$

for exact decomposition: $C_{HST}(U,V) = 0$ $\overline{F}(U,V) = 1$

How close is an approximation to the exact one?

Frobenius-norm based fidelity

$$\|A\|_{\mathrm{F}} = \left(\sum_{i=1}^m \sum_{j=1}^n |a_{ij}|^2
ight)^{rac{1}{2}}$$

The cost function of the optimization:

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$$f(U,V) = \frac{1}{2} \|V - U\|_F^2 = d - \operatorname{Re}\left[\operatorname{Tr}(U^{\dagger}V)\right]$$

The Fidelity:

(C)

$$\overline{F}_F(U,V) = 1 - \frac{d}{d+1} + \frac{1}{d(d+1)} \left(d - f(U,V)\right)^2$$

$$\overline{F}_F(U,V) \le \overline{F}(U,V)$$

"Best Approximate Quantum Compiling Problems"

Liam Madden (University of Colorado), Andrea Simonetto (IBM Research Ireland) arXiv:2106.05649



ACM Transactions on Quantum Computing • Accepted on June 2022 • https://doi.org/10.1145/3548693

 $\mathbf{U}(\mathbf{n}_{6},\bar{\mathbf{x}}) = (I_{2} \otimes U_{1}^{3} \otimes U_{2}^{3})(I_{2} \otimes C_{12}^{2})$

Quantum Fast Approximate Synthesis Tool (QFAST)



arXiV > quant-ph > arXiv:2003.04462

n-qubit unitary
$$\rightarrow \frac{n}{2}$$
-qubit unitaries $\rightarrow \frac{n}{4}$ -qubit unitaries

 \rightarrow 1 and 2-qubit unitaries

$$U(2^n) = \{ e^{i(\boldsymbol{\alpha} \cdot \boldsymbol{\sigma}^{\otimes n})} \mid \boldsymbol{\alpha} \in \mathbb{R}^{4^n} \}$$

$$\boldsymbol{\sigma}^{\otimes n} = \{\sigma_j \otimes \sigma_k \mid \sigma_j \in \boldsymbol{\sigma}, \sigma_k \in \boldsymbol{\sigma}^{\otimes n-1}\} \quad \boldsymbol{\sigma} = \{\sigma_i, \sigma_x, \sigma_y, \sigma_z\}$$



Adaptive quantum gate decomposition (SQUANDER)



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(C)

Expansion of controlled R_v rotations







Journals & Magazines > IEEE Transactions on Computer... > Volume: 38 Issue: 7 🔞

An Efficient Methodology for Mapping Quantum Circuits to the IBM QX Architectures



Alwin Zulehner (10); Alexandru Paler; Robert Wille (10) All Authors

- In the benchmark we tested the decomposition of 3, 4 and 5-qubit unitaries from online database containing series of circuits published as part of the Qiskit Developer Challenge, a public competition to design a better routing algorithm.
- quantum circuits of well known algorithms:
 - Grover search,
 - Quantum Fourier Transformation (QFT)
 - Quantum Approximate Optimization Algorithm (QAOA),
 - Quantum variational eigensolver (VQE)



https://github.com/iic-jku/ibm_qx_mapping

Gate synthesis benchmark



File name	Initial	QISKIT	SQUANDER		QFAST		QSEARCH	
	CNOT	CNOT	CNOT	$\overline{T}\left[s ight]$	CNOT	$\overline{T}\left[s ight]$	CNOT	$\overline{T}\left[s ight]$
4gt5_77	58	338	23	1293	26	332	-	-
4gt13_91	49	187	23	1296	25	732	48	2324
ham3_102	11	15	6	4.9	7	3.2	8	2.6
4gt5_76	46	529	24	1711	29	476	-	-
alu-v0_26	38	204	23	7900	42	912	29	9284
miller_11	23	18	8	7	9	5.4	10	4.5
rd32_v1_68	16	66	9	23.9	13	21.6	13	615
4mod5-v0_20	10	526	9	3650	17	166	16	14508
alu-v0_27	17	212	17	3452	30	674	34	3801
mod5mils_65	16	73	12	11162	20	405	-	-
ex-1_166	9	20	9	4.4	8	4.7	8	5.9
$decod24-v1_41$	38	130	20	2414	36	413	24	349
alu-v3_34	24	237	25	6090	37	1814	27	7834
3_17_13	17	23	7	6.5	9	4.2	9	4.3
4gt11_84	9	163	9	642	20	318	-	-
$decod24-v0_38$	23	48	14	62	23	58	15	285
4mod5-v0_19	16	75	13	701	21	375	-	-
4mod5-v1_22	11	168	9	962	13	52	17	82

gate fidelity: $\overline{F}_F = 1 - \varepsilon$ $\varepsilon \approx 10^{-9}$

Gate synthesis benchmark



File name	Initial	QISKIT	SQUANDER		QFAST		QSEARCH	
	CNOT	CNOT	CNOT	$\overline{T}\left[s\right]$	CNOT	$\overline{T}\left[s ight]$	CNOT	$\overline{T}\left[s ight]$
alu-v1_29	17	240	19	3820	33	801	-	-
alu-v1_28	18	331	19	2488	36	607	-	-
4mod5-v1_23	32	74	13	946	40	702	-	-
4mod5-v0_18	31	671	15	1134	31	266	-	-
rd32_270	36	522	14	893	27	627	-	-
rd32-v0_66	16	66	10	29	16	25	13	443
alu-v3_35	18	249	20	3655	31	1050	-	-
4gt13-v1_93	30	218	23	2408	38	466	33	21315
4mod5-v1_24	16	241	14	5081	33	210	52	3968
mod5d1_63	13	76	13	867	29	304	-	-
alu-v4_36	51	193	40	11090	49	2343	-	-
4gt11_82	18	419	15	883	22	698	19	1003
4gt5_75	38	259	25	7002	37	429	49	33246
alu-v2_33	17	358	17	2339	31	665	23	6520
4gt11_83	14	151	13	1994	15	98	19	1107
$decod24-v2_43$	22	46	9	93	19	44	17	1390
4gt13_92	30	161	24	1767	46	1830	-	-
alu-v4_37	18	276	18	3509	37	837	32	2142
mod5d2_64	25	129	14	846	26	104	16	256

gate fidelity: $\overline{F}_F = 1 - \varepsilon$ $\varepsilon \approx 10^{-9}$

Complexity analysis of the calculations



Number of qubits

The cost function of the optimization:

$$f(U,V) = \frac{1}{2} \|V - U\|_F^2 = d - \operatorname{Re}\left[\operatorname{Tr}(VU^{\dagger})\right]$$

Number of gates

The computational cost to evaluate VU^{\dagger} is M x 4ⁿ trial circuit to input quantum synthesize U program 2ⁿx2ⁿ

Gradient components also need to be calculated





Hardware accelerator



FPGA implementation of a quantum computer simulator

computational concurrency (on-chip multipliers used for multiplications)



number of supported qubits (converted into on-chip memory usage)

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reasonable trade-off
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computational accuracy (fixed point number representation, bitwidth)

- support for **arbitrary quantum circuit** composed of single qubit rotations and conditional two-qubit gates
- don't recompile the FPGA implementation when the gate structure is changed
- High level development framework of MAX





Data-flow implementation of a quantum computer simulator



Organize data into streams flowing through the chip

Computations: operations on the elements of a data stream



FPGA hardware + data-flow programming model = **Data-flow engine (DFE)**



DFE flavour of quantum gate operations



The elementary gate operations can be represented by sparse unitaries, mixing element pairs in the columns of V

Organizing the columns of V into a stream of data

DFE model of gate operations

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DFE implementatio of quantum gate operations







Complexity of a gate operation



Amplitude transformation $C_{\alpha} = U_{\alpha,0}C_0 + U_{\alpha,1}C_1$

2 complex multiplication and 1 complex addition

digital signal processing (DSP) units for multiplicatios have input ports:

18 bit x 27 bit \longrightarrow 32bit multiplications needs to be tiled

Karatsuba multiplication of 32-bit integers (W=16 bits)

$$A \times B = (a_1 2^W + a_0) (b_1 2^W + b_0) = a_1 b_1 2^{2W} + (a_1 b_0 + a_0 b_1) 2^W + a_0 b_0$$

3 multiplications instead of 4
and 5 additions

$$(a_0 + a_1) (b_0 + b_1) - a_0 b_0 - a_1 b_1$$

$$16 \times 16$$

$$(a_0 + a_1) (b_0 + b_1) - a_0 b_0 - a_1 b_1$$

Use Karatsuba strategy for complex multiplications as well

In total: $2 \times 3 \times 3 = 18$ multiplications $\rightarrow 18$ DSP units are needed







+ look-up-tables (LUTs)

DFE quantum computer simulator





- Chain up successive gate operations to increase computational concurrency
- Buffer the transformed unitary into the on-board memory (64 GB)



DFE quantum computer simulator

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 $6 \times 18 = 108$ gates on each Super Logic Region (SLR) 6 asynchronously operating gate blocks Each block contains 18 synchronously operating gate operations Xilinx Alveo U250 FPGA chips contain 4 SLRs In total 432 parallel gate operations on a single FPGA chip

Memory controller



SLR

DFE QC simulator performance



arithmetic operations per second: $18 \cdot 4 \cdot N_{G,chain} \cdot f_{\text{DFE}}$ (excluding all integer logic from the count)

equvalent to 2.72 TOPS

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DFE vs CPU performance







Gate synthesis benchmark Quantum Information National Laboratory HUNGARY



Circuit name	n	IBM QX (39)		QISKIT (40)		SQUANDER (41)			comp.
		CX	D	CX	D	CX	D	f	rate
4gt12-v0_87	6	112	131	625	1146	47	73	0.0028	93.6%
4gt12-v0_88	6	86	108	853	1647	44	71	0.0072	95.7%
4mod5-bdd_287	7	31	41	1037	1825	26	41	0.012	97.8%
alu-bdd_288	7	38	48	224	408	30	35	0.0038	91.4%
C17_204	7	205	253	2992	5915	104	133	0.0042	97.8%
ex2_227	7	275	355	2852	5554	133	161	0.0128	97.1%
majority_239	7	267	344	4024	7950	143	175	0.0127	97.8%
rd53_131	7	200	261	6538	12320	93	119	0.0129	99.0%
rd53_135	8	134	159	26126	50436	120	147	0.0195	99.7%
rd53_138	8	60	56	18567	35172	87	117	0.061	99.7%
cm82a_208	8	283	337	11246	22284	86	67	0.0129	99.7%
con1_216	9	415	508	55822	109798	205	229	0.118	99.8%

 $\overline{F}_F = 1 - \varepsilon$ $\varepsilon \approx 10^{-4}$







Further improvements to the compiler COLD Quantum Information National Laboratory HUNGARY

Construct the initial gate structure

interations to compress the initial gate structure

for 8-9 qubits

VS

• takes a 2-7 days

takes a 1-30 minutes (each iteration)

- In both cases we need to solve quite similar optimization problem.
- Why is the big difference in the execution time?
- 2 Scale up the compiler to circuits with many qubits
 - Reorganize circuit to exhibit large 6-10 qubit subblocks that can be optimized



Iterative Gate Compression



• Try to remove a random 2-qubit block



- Try to solve the optimization problem for the reduced circuit
- The reduced initial parameter set is derived from the previous solution

 $p_1, p_2, p_3, p_4, p_5, p_6, p_7, p_8, p_9, p_{10}, p_{11}, p_{12}, p_{13}, p_{14}, p_{15}, p_{16}, p_{17}, p_{18}, p_{19}, p_{20}, \dots$

- Solve the optimization problem in 1-30 minutes
- A well chosen initial parameter set (i.e. correlations between the qubits) is crucial to speed up the quantum compilation



How to chose the initial parameter set?

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VXM: int8, int16, int32, uint8, uint16, uint32, float16, float32, bool8, bool16, bool32

MXM: int8 x int8 \rightarrow int32, float16 x float16 \rightarrow float32



Concept of Groq QC simulator



gate

kernel



unitary transformation:



- In principle 50% more perfromance by Groq compared to DFE
- The preformance can be keep up to 11 qubits (on DFE the limit is 9 qubits)

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Conclusions and outlook

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We have designed a DFE based QC simulator to speed up the gate synthesis process up to 9 qubit circuits.

Aiming to reduce the execution time by:

- Predict intial parameter set with machine learning (achieve competitive execution time with deterministic tools)
- Scale up the decomposition for circuits with more qubits (transform the circuit with gate identities, optimize 6-9 qubit blocks)





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