

HPCA 2025 Tutorial

Topic 3. MorphQPV: Exploiting Isomorphism in Quantum Programs to Facilitate Confident Verification







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https://janusq.github.io/HPCA_2025_Tutorial/

Outline of Presentation





- Background and Challenges
- Overview of MorphQPV
- Assertion Statement and Validation
- Experiment
- API of MorphQPV

Ensure Program Is Correct?

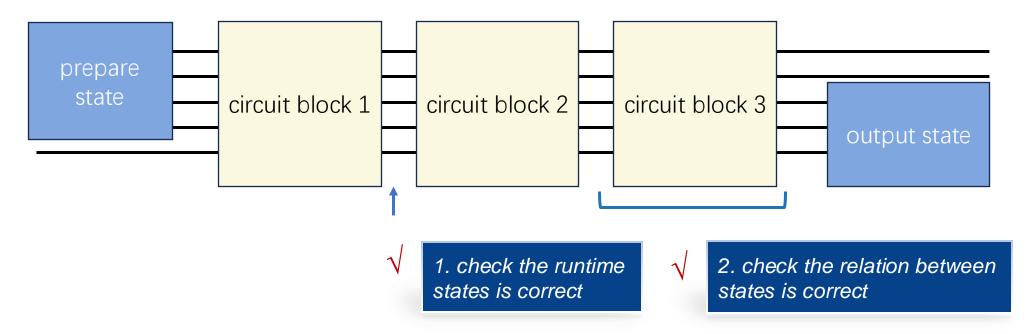




Quantum program verification is a fundamental theoretical method for reliable quantum computing, which aims to ensure that quantum programs have correct behavior and achieve desired results during execution.

Process to check the program is correct

- 1. Check the relationship between the states in each stage
- 2. There may be mid-measurement or feedback



Described as **an assertion**, which is defined as a predicate. The predicate is expected to be true if there are no bugs.

Ensure Program Is Correct?





Quantum program verification is a fundamental theoretical method for reliable quantum computing, which aims to ensure that quantum programs have correct behavior and achieve desired results during execution.

Process to check the program is correct

- 1. Check the relationship between the states in each stage
- 2. There may be mid-measurement or feedback

assert $a \ge 0$

$$b = \sqrt{a}$$

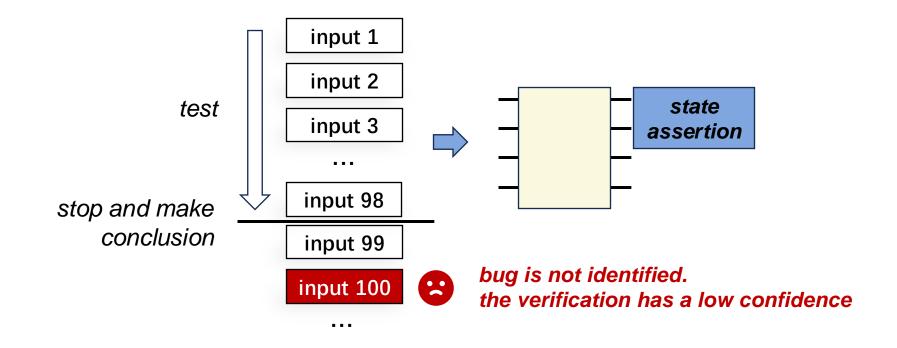
Quantum assertion is similar to the classical assertion, while the verified qubits usually stay in the superposition state.

Confidence of the Verification





Confidence of the verification: The probability that the correctness holds for all inputs.



The ability to generalize the test inputs to the whole space is necessary to ensure high confidence

Limitation of Current Quantum Assertion Methods

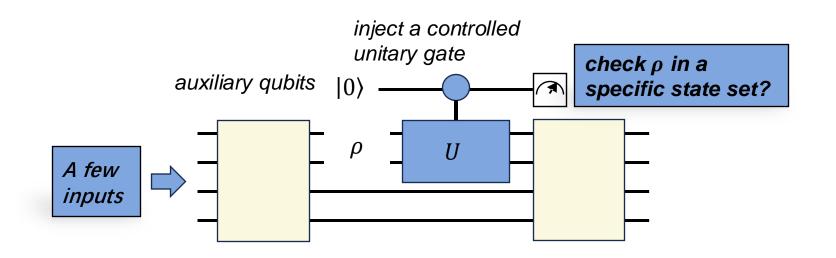




	Statement	Analysis method	Input coverage	Theorem
A good assertion validation method	Specify the relation between states	Efficiently characterize the relation	The whole input space	 Complexity is up-bounded Theorem guarantee to high confidence

An example

Ji Liu, et al. Quantum circuits for dynamic runtime assertions in quantum computation. ASPLOS, 2020



Limitation of Current Quantum Assertion Methods

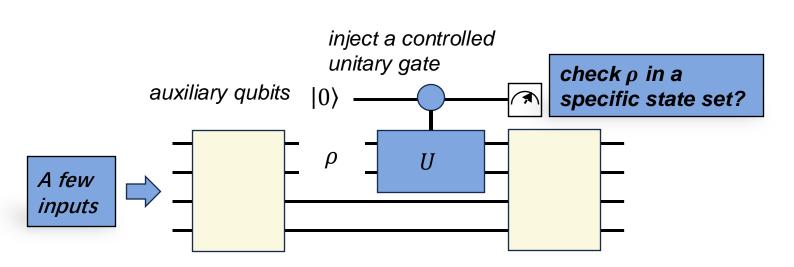




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Expressiveness?

Only support state in state equal operation.

Efficient?

Numerous gates to synthesize unitary gates.

High confidence?

Cover a few inputs. Lack confidence guarantee.

Limitation of Current Quantum Assertion Methods

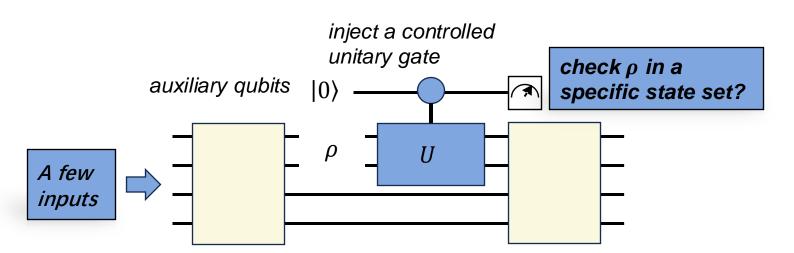




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An example

Ji Liu, et al. Quantum circuits for dynamic runtime assertio



Similar problems in current works, including Liu, et al. OOPSLA 2020, Huang, et al. ISCA, 2019

Expressiveness?

Only support state in state equal operation.

Efficient?

Numerous gates to synthesize unitary gates.

High confidence?

Cover a few inputs. Lack confidence guarantee.

Reason of Limitations



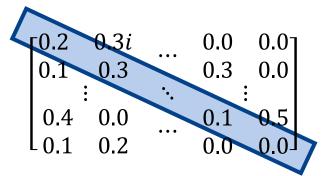


Quantum collapse leads to loss of information.



N-qubit state

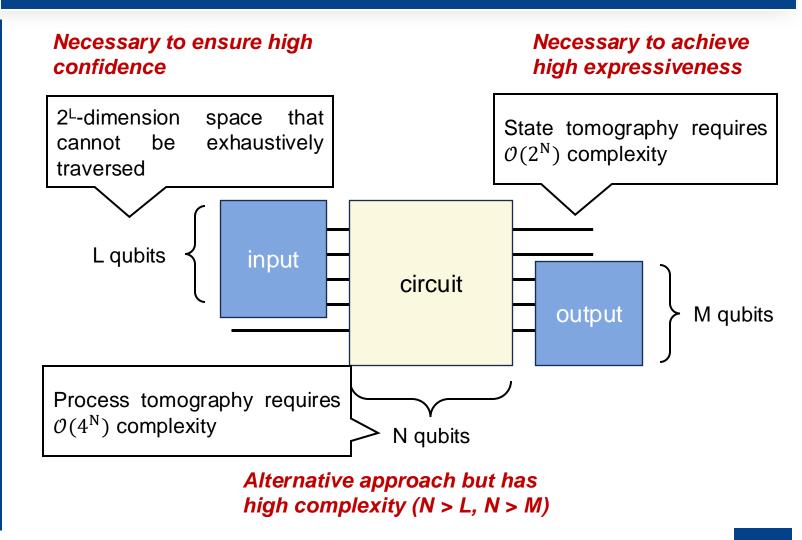
=



 $2^{N} \times 2^{N}$ -size density matrix

Only diagonal elements are obtained by measurements

Lack an efficient method to characterize the mapping from input to output



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write a quantum program

input q[0,1];

x q[2,3,4];
cz q[1],q[4];
t1 q[1,2];
x q[2,3,4];
h q[1];
t2 q[0];



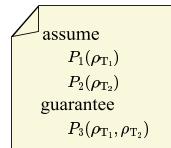
label states by a tracepoint pragma

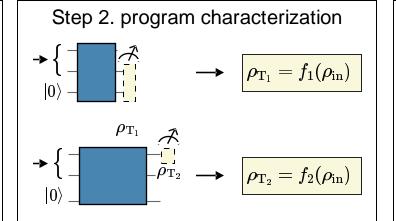
Object: enable input-independent assertion

Object: minimize the number of program execution

Object: apply global search to counter example

Step 1. assertion statement



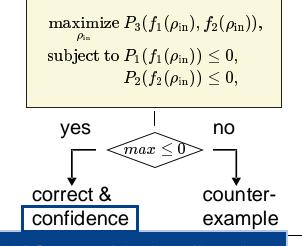


compare

specify the **expected relation** between the tracepoint states

characterize the **natural relation** by building approximation functions

Step 3. assertion validation



Object: guide the allocation of computation resource





write a quantum program

```
input q[0,1];

x q[2,3,4];

cz q[1],q[4];

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```

label states by a tracepoint pragma





write a quantum program

input q[0,1];

x q[2,3,4];
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x q[2,3,4];
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t2 q[0];

label states by a tracepoint pragma

Object: enable input-independent assertion



Step 1. assertion statement

assume $P_1(
ho_{ ext{T}_1})$ $P_2(
ho_{ ext{T}_2})$ guarantee $P_3(
ho_{ ext{T}_1},
ho_{ ext{T}_2})$

specify the **expected relation** between the tracepoint states





write a quantum program

input q[0,1];

x q[2,3,4];
cz q[1],q[4];

t1 q[1,2];
x q[2,3,4];
h q[1];

t2 q[0];

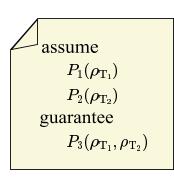
label states by a tracepoint pragma

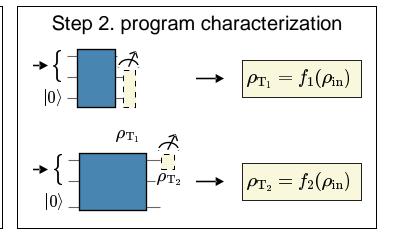
Object: enable input-independent assertion

Object: minimize the number of program execution



Step 1. assertion statement





specify the **expected relation** between the tracepoint states

characterize the **natural relation** by building approximation functions





write a quantum program

input q[0,1];

x q[2,3,4];
cz q[1],q[4];
t1 q[1,2];
x q[2,3,4];
h q[1];
t2 q[0];

label states by a tracepoint pragma

Object: enable input-independent assertion

Object: minimize the number of program execution

Object: apply global search to counter example



Step 1. assertion statement

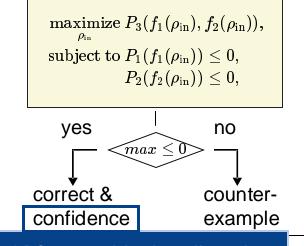
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ho_{ ext{T}_2})$

compare

specify the **expected relation** between the tracepoint states

characterize the **natural relation** by building approximation functions

Step 3. assertion validation



Object: guide the allocation of computation resource

Outline of Presentation

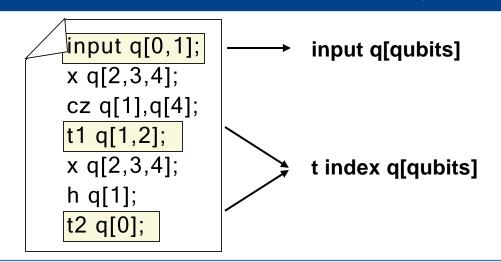


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Assertion Statement



1. Label the asserted state by tracepoint pragma



2. Use assume-guarantee assertion to specify the relation between states

assume:

$$P_1(\rho_{T1})$$
, $P_2(\rho_{T2})$

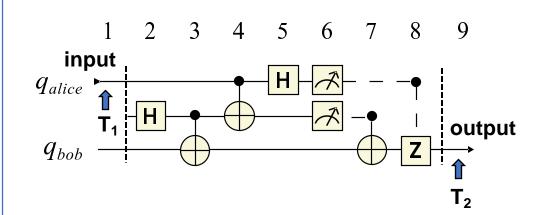
guarantee:

$$P_3(\rho_{T1},\rho_{T2})$$

If $P_1 \le 0$ and $P_2 \le 0$ then $P_3 \le 0$ \Rightarrow assertion is correct

If $P_1 \le 0$ and $P_2 \le 0$ then $P_3 > 0$ \Rightarrow assertion fails

An example: Quantum teleportation



input state should equal output state

assume:

$$P_1(\rho_{T1}) = \|\rho_{T1}\rho_{T1}^{\dagger} - \rho_{T1}\|,$$

$$P_2(\rho_{T2}) = \|\rho_{T2}\rho_{T2}^{\dagger} - \rho_{T2}\|,$$

guarantee:

$$P_3(\rho_{T1}, \rho_{T2}) = \|\rho_{T1} - \rho_{T2}\|$$

Isomorphism-based Characterization





Isomorphism

A structure-preserving mapping $\mathbb{R}_x \to \mathbb{R}_y$ between two spaces of the same type that can be retraced by an inverse mapping.

Example of isomorphism

$$x + 1 = y$$

$$\text{inverse}$$

$$y - 1 = x$$

Quantum evolution is isomorphism

$$U\rho U^{-1}=\rho'$$

$$\text{inverse}$$

$$U^{-1}\rho' U=\rho$$

Feature of isomorphism

additivity:
$$f(u + v) = f(u) + f(v)$$

homogeneity:
$$f(c u) = c f(u)$$

$$f(\sum_i c_i u_i) = \sum_i c_i f(u_i)$$



also has the feature

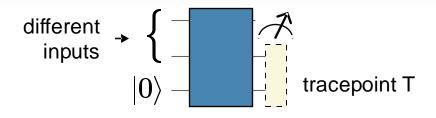
inspire us to generalize the information obtained from individual input into a broader input space

Isomorphism-based Characterization





Step 1: sample inputs



Inputs are orthogonal and prepared by the Clifford group.

Based on "Sergey Bravyi and Dmitri Maslov. Hadamard-free circuits expose the structure of the clifford group. IEEE Transactions on Information Theory, 2021".

Record inputs and state at tracepoint as $\langle \sigma_{input,i}, \sigma_{T,i} \rangle$ pairs.

Input state	Tracepoint state		
$\sigma_{ m input,1}$	$\sigma_{ m T,1}$	Obtained by	
$\sigma_{ m input,2}$	$\sigma_{ m T,2}$	tomography	
$\sigma_{ ext{input},N_{ ext{sample}}}$	$\sigma_{\mathrm{T},N_{\mathrm{sample}}}$		

Step 2: construct approximation function $f(\rho_{\mathrm{input}}) = \rho_{\mathrm{T}}$

The function is computed in two steps:

1. For input ρ_{input} , it first approximates the ρ_{input} to the linear combination of sampled inputs $\sigma_{inpu,i}$

$$\rho_{\text{input}} = \sum_{i} \alpha_{i} \sigma_{\text{input,i}}$$

$$\begin{cases}
\{\alpha_{i}\} \text{ is real parameters.}
\end{cases}$$

2. It then outputs tracepoint state:

$$\rho_{\mathrm{T}} = \sum_{i} \alpha_{i} \, \, \sigma_{\mathrm{T,i}}$$

Based on the additivity and homogeneity of isomorphism

$$f(\sum_{i} c_{i} u_{i}) = \sum_{i} c_{i} f(u_{i})$$

Accuracy of Characterization

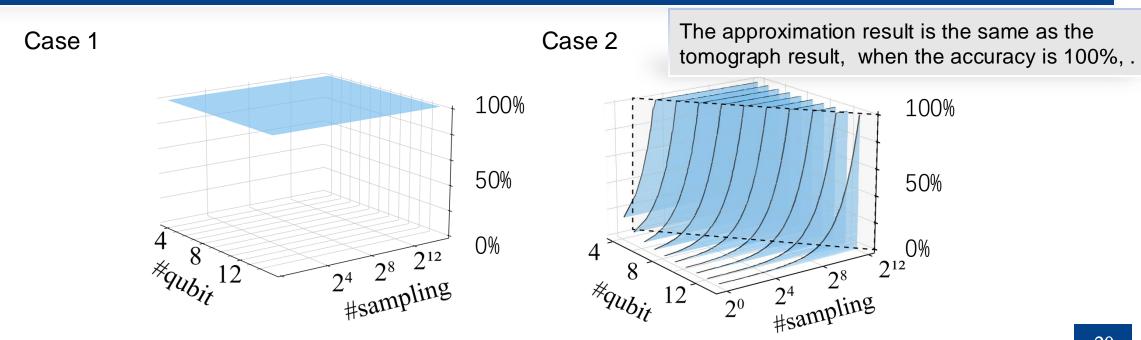




Theorem 1 (Approximation Accuracy)

- Case1: For inputs that can be accurately represented by linear combination of sampled inputs, the accuracy is 100%.
- Case2: For inputs with eigenstates that cannot be represented the linear combination of sampled inputs, the average accuracy is $\frac{N_{\text{sample}}}{2^{N_{input}}} \times 100\%$

Example: Approximation accuracy in the quantum teleportation programs with different number of qubits and sampled inputs.







Assertion

assume:

$$P_1(\rho_{T1})$$
 , $P_2(\rho_{T2})$

assume-guarantee:

$$P_3(\rho_{T1},\rho_{T2})$$

Maximization problem

$$\max_{\rho_{\text{input}}} P_3(\rho_{T1}, \rho_{T2}),$$

subject to $P_1(\rho_{T1}) \leq 0$,

$$P_2(\rho_{T2}) \le 0.$$

Maximization problem

$$\max_{\{\alpha_i\}} P_3(f_1(\{\alpha_i\}), f_2(\{\alpha_i\})),$$

subject to
$$P_1(f_1(\{\alpha_i\})) \leq 0$$
,

$$P_2(f_2(\{\alpha_i\})) \le 0.$$



Solver by solver, e.g. quadratic programming

Characterization

$$f_1(\rho_{\text{input}}) = \rho_{\text{T}1}$$

$$f_2(\rho_{\rm input}) = \rho_{\rm T2}$$



$$f_1(\{\alpha_i\}) = \rho_{\mathrm{T}1}$$

$$f_2(\{\alpha_i\}) = \rho_{\mathrm{T2}}$$

Approximation parameters

Validation

if $\max P_3 \leq 0$:

assertion is true

else:

assertion is false





Assertion

assume:

$$P_1(
ho_{T1})$$
 , $P_2(
ho_{T2})$

assume-guarantee:

$$P_3(\rho_{T1},\rho_{T2})$$



$$\max_{\rho_{\rm input}} P_3(\rho_{T1}, \rho_{T2}),$$

subject to $P_1(\rho_{T1}) \leq 0$,

$$P_2(\rho_{T2}) \le 0.$$





Assertion

assume:

$$P_1(\rho_{T1}), P_2(\rho_{T2})$$

assume-guarantee:

$$P_3(\rho_{T1},\rho_{T2})$$



Maximization problem

$$\max_{\rho_{\text{inpu}t}} P_3(\rho_{T1}, \rho_{T2}),$$

subject to $P_1(\rho_{T1}) \leq 0$,

$$P_2(\rho_{T2}) \le 0.$$

Characterization

$$f_1(\rho_{\rm input}) = \rho_{\rm T1}$$

$$f_2(\rho_{\rm input}) = \rho_{\rm T2}$$



$$f_1(\{\alpha_i\}) = \rho_{\mathrm{T}1}$$

$$f_2(\{\alpha_i\}) = \rho_{T2}$$

Approximation parameters





Assertion

assume:

$$P_1(\rho_{T1})$$
 , $P_2(\rho_{T2})$

assume-guarantee:

$$P_3(\rho_{T1},\rho_{T2})$$

Maximization problem

$$\max_{\rho_{\text{input}}} P_3(\rho_{T1}, \rho_{T2}),$$

subject to $P_1(\rho_{T1}) \leq 0$,

$$P_2(\rho_{T2}) \le 0.$$

Maximization problem

$$\max_{\{\alpha_i\}} P_3(f_1(\{\alpha_i\}), f_2(\{\alpha_i\})),$$

subject to
$$P_1(f_1(\{\alpha_i\})) \leq 0$$
,

$$P_2(f_2(\{\alpha_i\})) \le 0.$$



Solver by solver, e.g. quadratic programming

Characterization

$$f_1(\rho_{\text{input}}) = \rho_{\text{T}1}$$

$$f_2(\rho_{\rm input}) = \rho_{\rm T2}$$



$$f_1(\{\alpha_i\}) = \rho_{\mathrm{T}1}$$

$$f_2(\{\alpha_i\}) = \rho_{\mathrm{T2}}$$

Approximation parameters

Validation

if $\max P_3 \leq 0$:

assertion is true

else:

assertion is false

Confidence of Validation



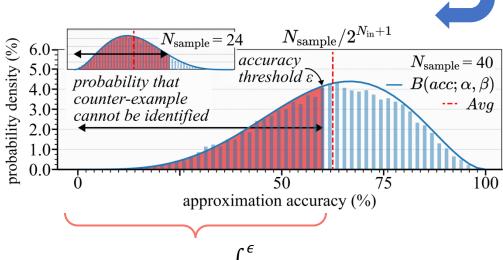


confidence = P(the correctness holds for all inputs) = 1 - P(counter-example exists but is not identified)

P(counter-example exists but is not identified) = P(accuracy of counter – example $\triangleleft \epsilon$)



accuracy threshold to discriminate error



P(accuracy
$$< \epsilon$$
) = $\int_0^{\epsilon} B(x; \beta_1, \beta_2)$

Accuracies follow Beta distribution $B(\beta_1, \beta_2)$ β_1 , β_2 can obtained by fitting some test inputs

Theorem 2 (Confidence)

When the program only has one counterexample lower-bound

confidence =
$$1 - P(accuracy < \epsilon)$$

When the program only has $N_{c-\rho}$ counterexamples

confidence =
$$1 - P(accuracy < \epsilon)^{N_{c-e}}$$

Accuracy and confidence linearly increase as the number of sampled inputs grows

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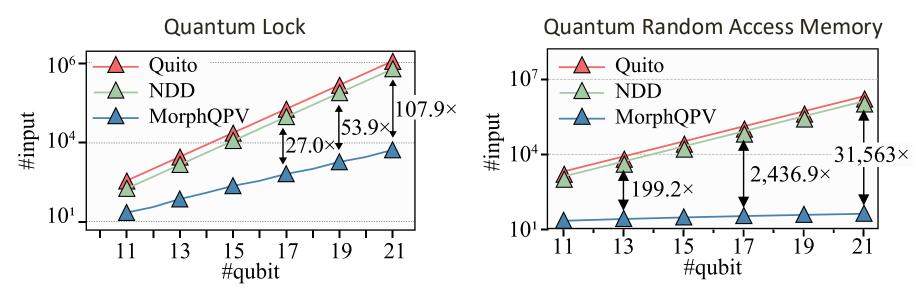
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Comparison to Prior Works





	Huang, et al. ISCA'19	Li,et al. OOPSLA'20	Liu, et al. ASPLOS'20	Feng, et al. ASPLOS'23	MorphQPV
Verified Object	Probability distribution	Mixed state	Mixed state	Mixed state	Mixed state & Evolution
Comparison	Part	Equal & In	Equal & In	Equal & In	Full
Interpretability	Part	No	No	No	Full
Feedback	No	No	No	No	Full



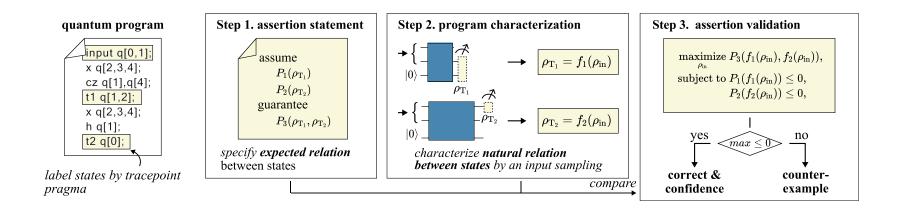
The number of test inputs in debugging the programs with different number of qubits

Conclusion



- 1. Limitations of prior assertion works: low confidence, low expressiveness, and high overhead
- 2. Three-step verification of MorphQPV: statement, characterization, and validation
- 3. Two theorems: upper-bound complexity of verification and lower-bound of confidence
- 4. Contents that are not mentioned in the presentation:
 - Proof of theorems.
 - Further optimization to minimize the overhead_o
 - detailed comparison with prior works.

Please refer to the paper.



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API of MorphQPV





File:

- JanusQ/examples/ipynb/3_1_verify_quantum_program.ipynb
- https://janusq.github.io/tutorials/demo/ 3_1_verify_quantum_program

```
from janusq.verification.morphqpv import MorphQC,Config
                                   from janusq.verification.morphqpv import IsPure, Equal, NotEqual
                                   myconfig = Config()
myconfig.solver = 'sgd'
           configure solver
                                   with MorphQC(config=myconfig) as morphQC:
                                      morphQC.x([1,3])
                                      morphQC.y([0,1,2])
                                      for i in range(4):
                                        morphQC.cnot([i, i+1])
   assertion statement and
                                      morphQC.s([0,2,4])
validation in quantum circuit
                                      morphQC.add_tracepoint(2,4)
                                      morphQC.assume(1,lsPure())
                                      morphQC.guarantee([1,2],Equal())
                                      morphQC.guarantee([0,1],NotEqual())
```



Thanks for listening

MorphQPV: Exploiting Isomorphism in Quantum Programs to Facilitate Confident Verification

Siwei Tan*, Debin Xiang*, Liqiang Lu†, Junlin Lu, Qiuping Jiang, Mingshuai Chen, and Jianwei Yin†