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# WELCOME TO TUTORIAL

## Session 3 JanusQ-CT: A Framework for Analyzing Quantum Circuit by Extracting Contextual and Topological Features



JanusQ  
Cloud



<https://janusq.github.io/tutorials/>

College of Computer Science and  
Technology,  
Zhejiang University

# Outline of Presentation

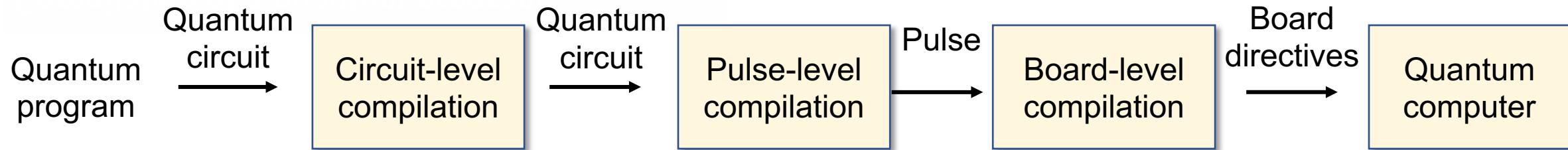
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- **Background and challenges**
- Janus-CT overview
- Upstream model: Circuit feature extraction
- Downstream model 1: Circuit fidelity prediction
- Downstream model 2: Unitary decomposition



## Compilation of a quantum program



### Circuit-level compilation:

- **Input:** quantum circuit

**Output:** Quantum circuit that satisfies the constraints

### Pulse-level compilation:

- **Input:** quantum circuit

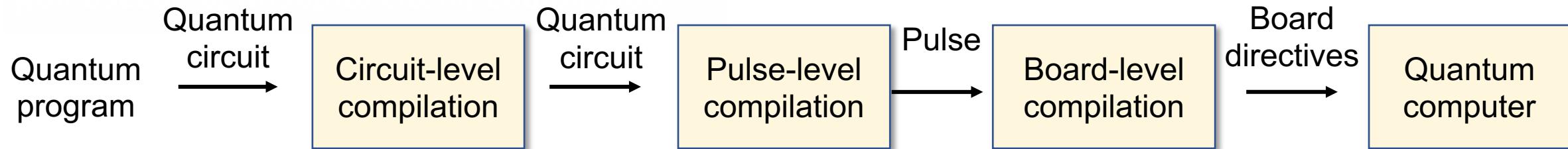
**Output:** Pulses received by qubits

### Board-level compilation:

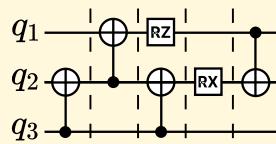
- **Input:** pulses

**Output:** Board directives

## Key passes of quantum circuit compilation



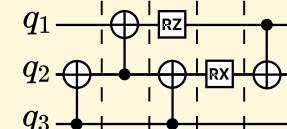
### Pass 1: Unitary decomposition



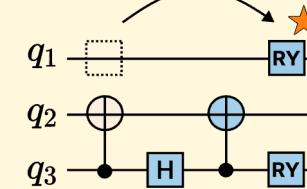
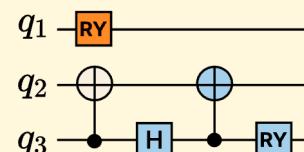
compute  
**small complexity**

$$\begin{bmatrix} 0 & e^{i-\pi/2} & 0 & 0 \\ 0 & 0 & 0 & e^{i-\pi/2} \\ e^{i\pi/2} & 0 & 0 & 0 \\ 0 & 0 & 0 & e^{i\pi/2} \end{bmatrix}$$

$$\begin{bmatrix} 0 & e^{i-\pi/2} & 0 & 0 \\ 0 & 0 & 0 & e^{i-\pi/2} \\ e^{i\pi/2} & 0 & 0 & 0 \\ 0 & 0 & 0 & e^{i\pi/2} \end{bmatrix} \xrightarrow{\text{decompose}} \text{large complexity}$$



### Pass 2: Fidelity prediction and optimization

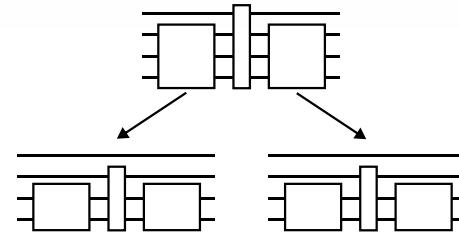


**Optimize the noise while keeping the equivalence of circuits**

# Challenges of Unitary Decomposition

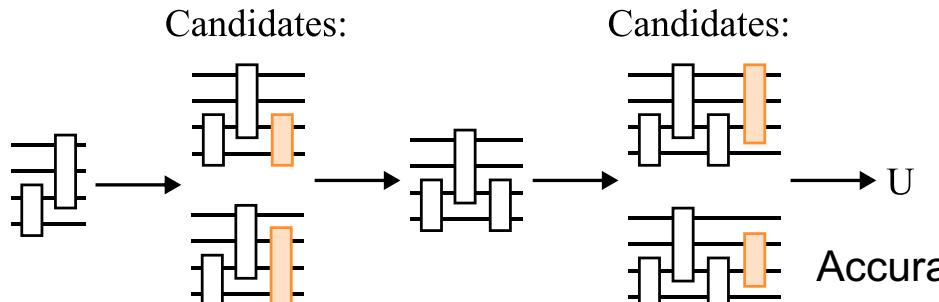


## Unitary decomposition



Fast but  
Leads to numerous redundant gates

### Template-based method



Accurate but  
Lacks a heuristic to prune candidate space.

### Search-based method

Category		Template-based	Search-based	
Method		CCD [1]	QSD [2]	QFAST [3] Squander [4]
Time		3.6 s	2.1 s	<b>511.2 h</b>
#Gate		<b>3,592</b>	<b>3,817</b>	806

$O(4^N)$  #Gate

$O(4^N)$  Time

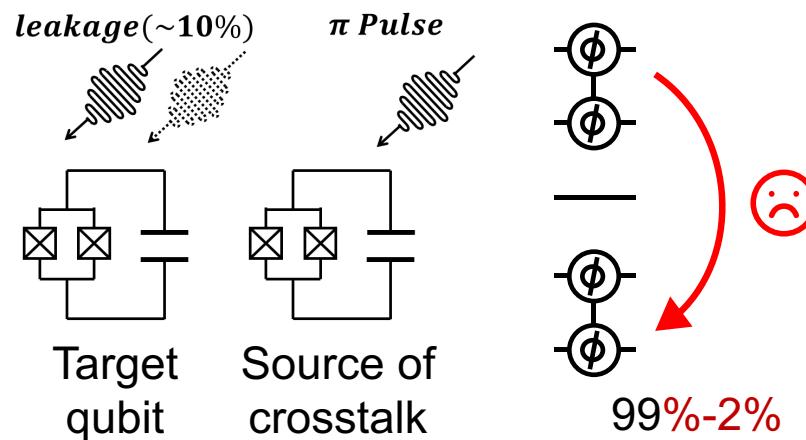
5-qubit unitary decomposition

- [1] R. Iten, et al. PRA. 2016
- [2] V. Shende, et al. ASP-DAC. 2005
- [3] E. Younis, et al. QCE. 2021.
- [4] P. Rakyta , et al. Quantum, 2022

# Challenges



## Fidelity prediction



Related to  
the circuit  
structure

Category	RB [5]	XEB [6]	Cycle bench. [7]	Noisy simulat. [8]
Gate-independent error	✓	✓	✓	✓
Crosstalk, Pulse distortion	✗	✗	✓	✓
Inaccuracy (IBMQ Manila )	<b>4-28%</b>	<b>3-36%</b>	<b>2-12%</b>	<b>3-17%</b>

**Fidelity prediction**

**Not one-shot**

[5] E. Knill , et al. D. PRA. 2008.

[6] F. Arute, et al. Nature. 2019

[7] A. Erhard, et al. Nature communications. 2019

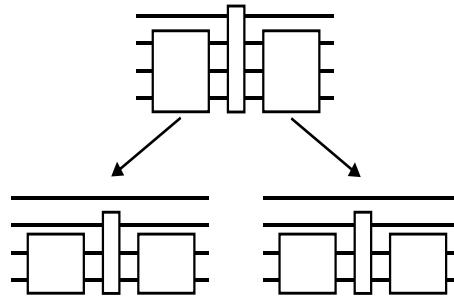
[8] Isakov, et al ArXiv. 2021.

# Current Compilation Methods



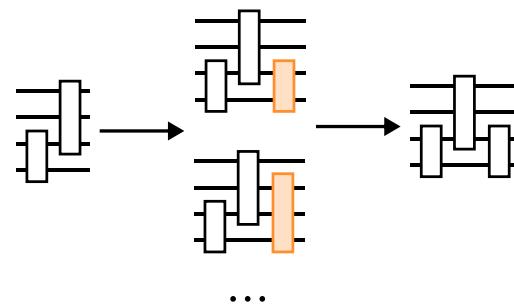
## Unitary decomposition

### Template-based method



**Fast by not accurate:**  
10-qubit unitary ->  
20,000 gate

### Search-based method



**Accurate but slow:**  
10-qubit unitary->  
one year

## Fidelity prediction

Method	Independent noise	Dependent noise	Inaccuracy
RB	✓	✗	4-28%
XEB	✓	✗	3-36%
CB	✓	✓	2-12%
Noisy Simulat.	✓	✓	3-17%

**Fast by inaccurate:**  
cannot model dependent noise

**Accurate but slow:**  
require repeated executions

**They face a trade-off between the efficiency and accuracy**

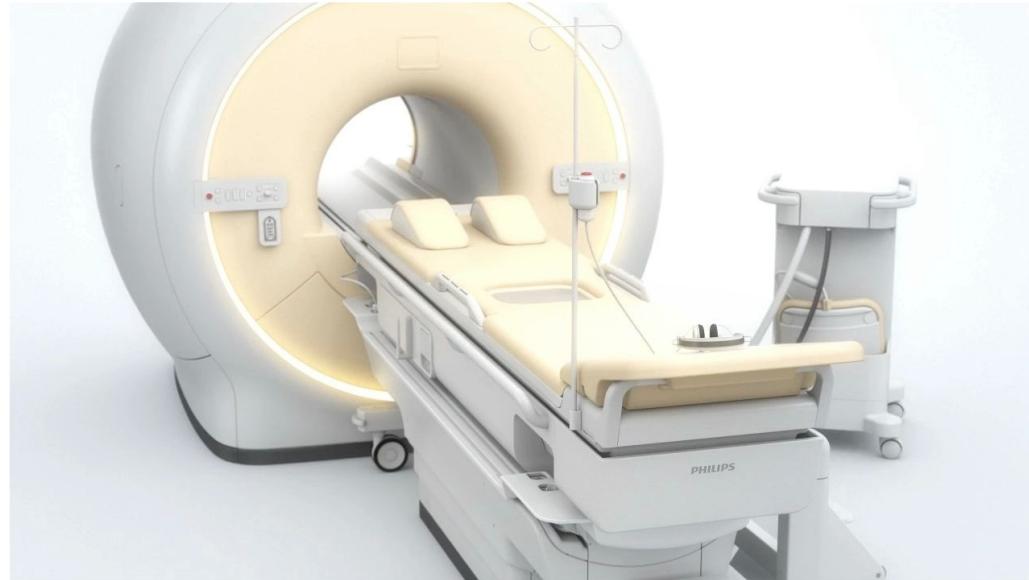
# Outline of Presentation

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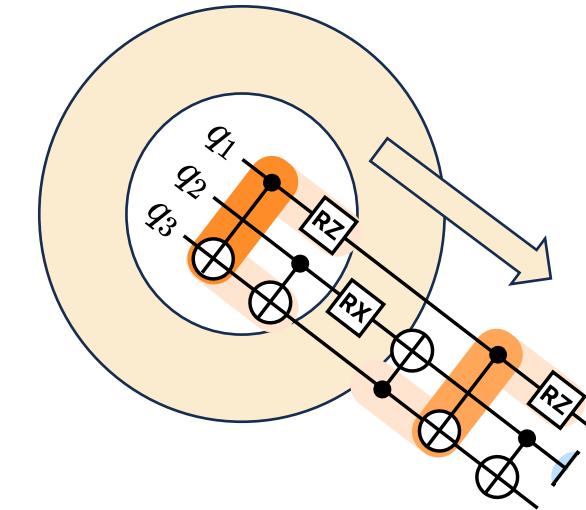


- Background and challenges
- **Janus-CT overview**
- Upstream model: Circuit feature extraction
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## Origin of the name

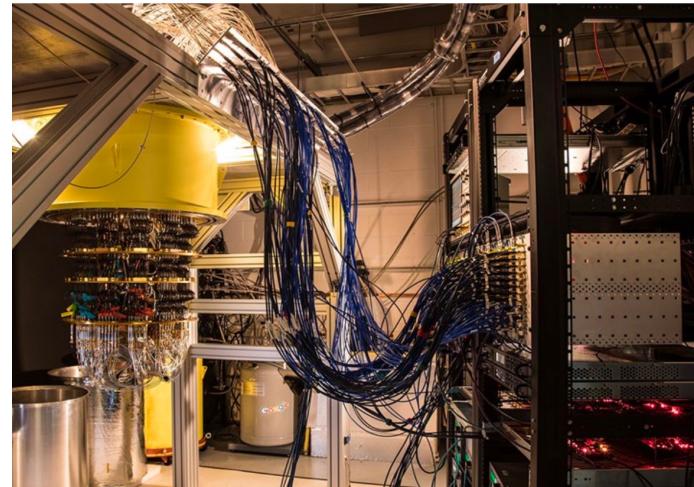


Computerized Tomography

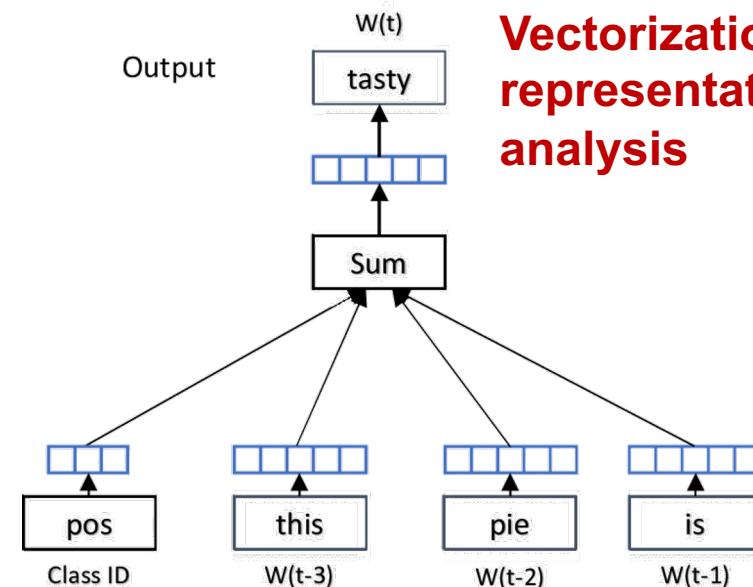


Analyzing Quantum Circuit by  
Contextual and Topological Features

## Solution: Implement circuit topology and context-aware gate vectorization



Quantum circuits are implemented via pulses. There are **interactions between wirings of qubits**.



Context extraction is common in **natural language processing (NLP)** and **classical program analysis**

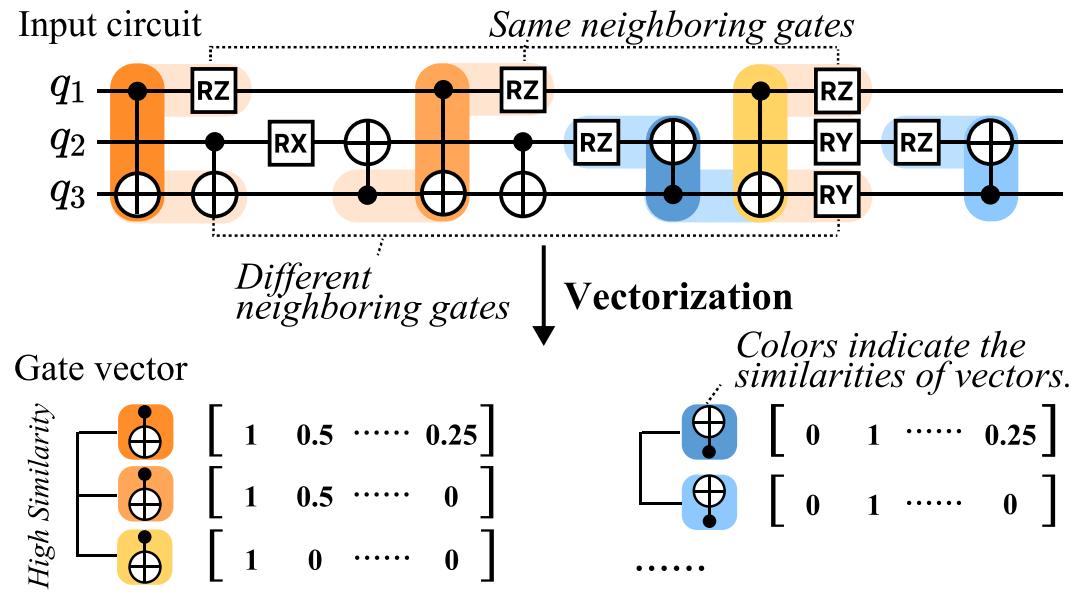
**Vectorization is a more efficient representation for further analysis**

Quantum	NLP
Fidelity prediction	grammar analysis
Circuit generation	Test generation

Quantum program analysis and NLP have similar tasks

Each model is one-shot generated

## Upstream Model:



## Downstream Model:

### Circuit Fidelity Prediction

a) Circuit fidelity prediction

$$E_{gate} = W^\top v_{gate}$$

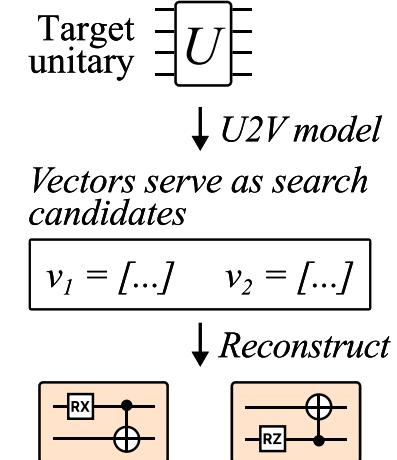
.1%      .5%      .3%

$F_{circuit} = 97.87\%$

b) Compilation- and calibration-level optimizations

More tasks: gate cancellation, bug detection ...

### Unitary Decomposition



Random walk

Vectorization

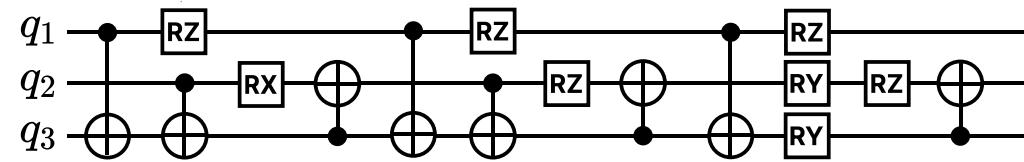
Fidelity prediction

or

Unitary decomposition

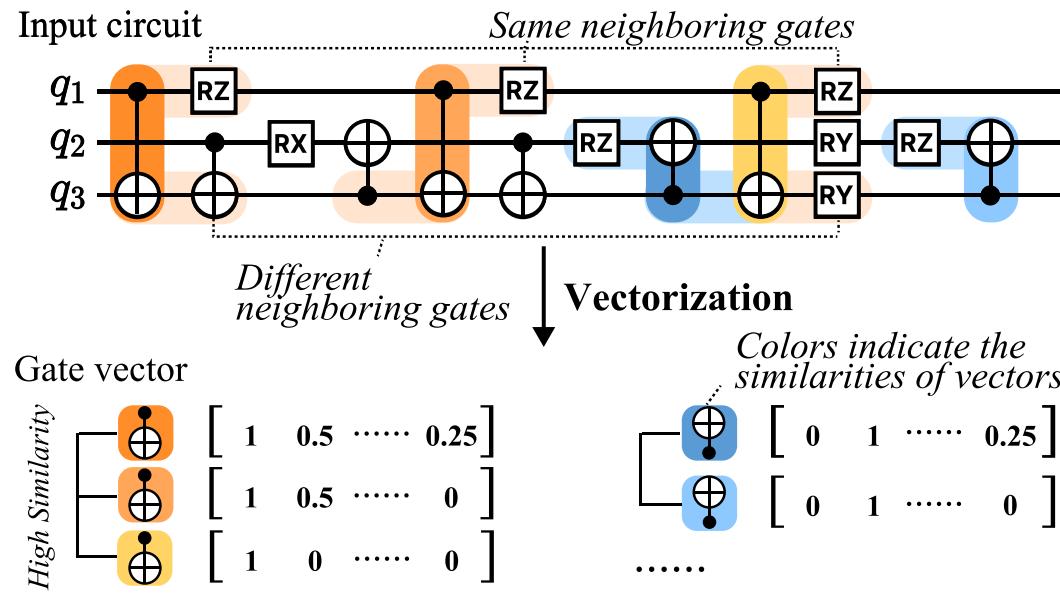
## Upstream Model:

Input circuit



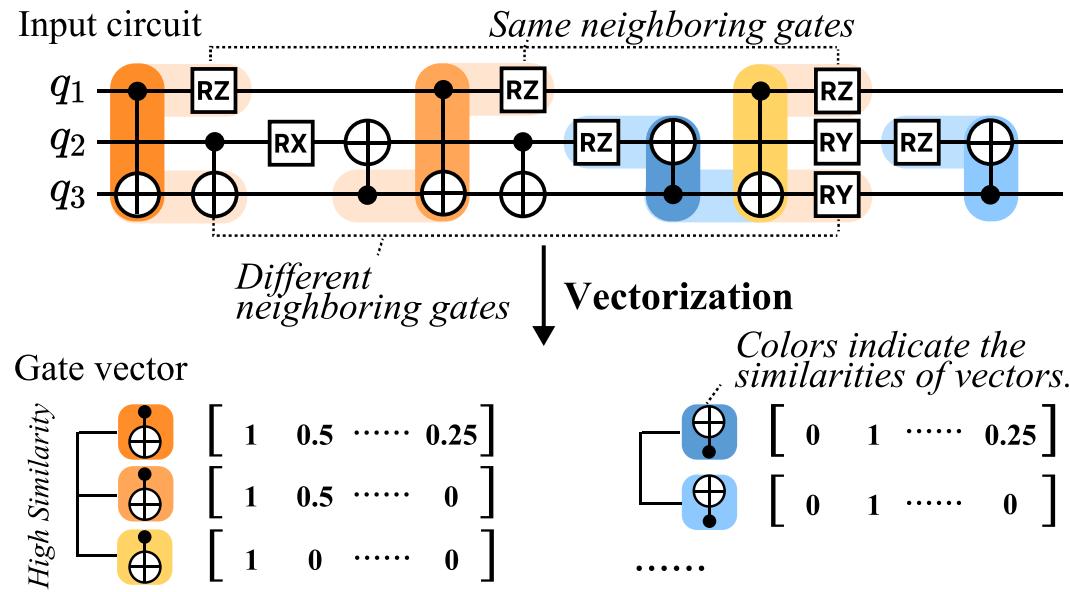
Random walk

## Upstream Model:



Random walk → Vectorization

## Upstream Model:



## Downstream Model:

### Circuit Fidelity Prediction

a) Circuit fidelity prediction

$$E_{gate} = W^\top v_{gate}$$

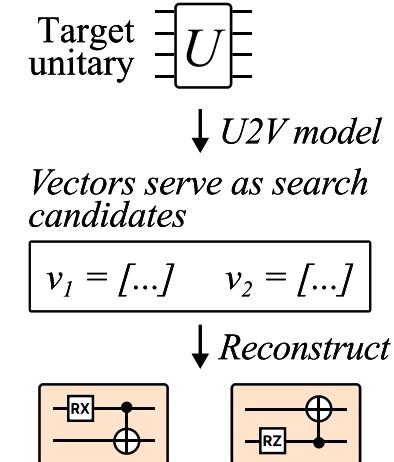
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### Unitary Decomposition



Random walk

Vectorization

Fidelity prediction

or

Unitary decomposition

# Outline of Presentation

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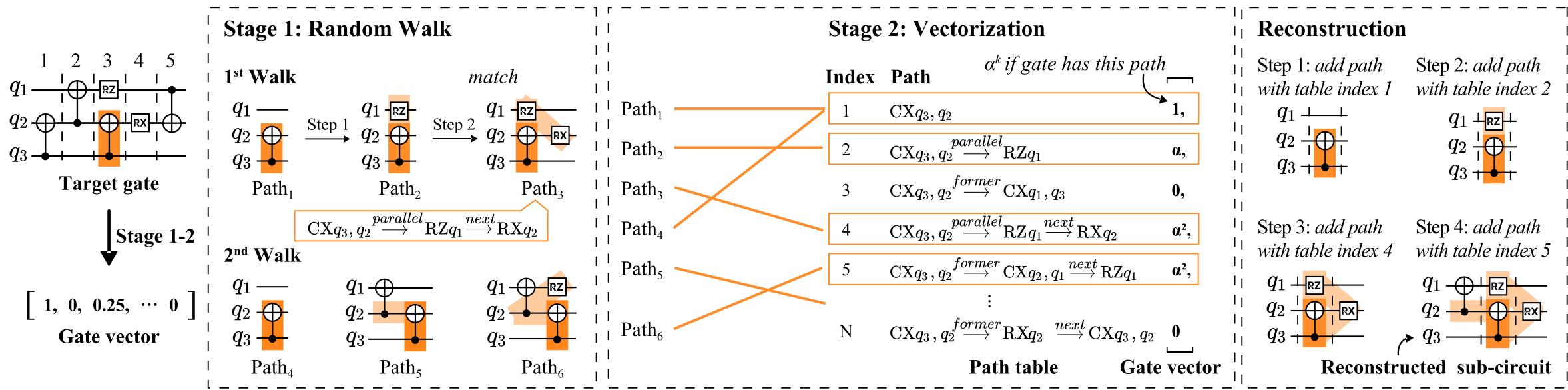


- Background and challenges
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- **Upstream model: Circuit feature extraction**
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# Upstream Model: Circuit Feature Extraction



## Two-step vectorization flow



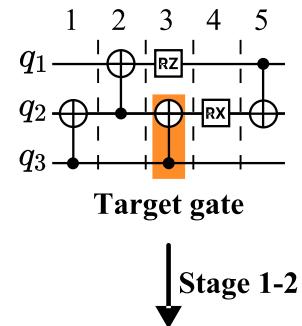
# Upstream Model: Circuit Feature Extraction



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## Two-step vectorization flow

For each gate



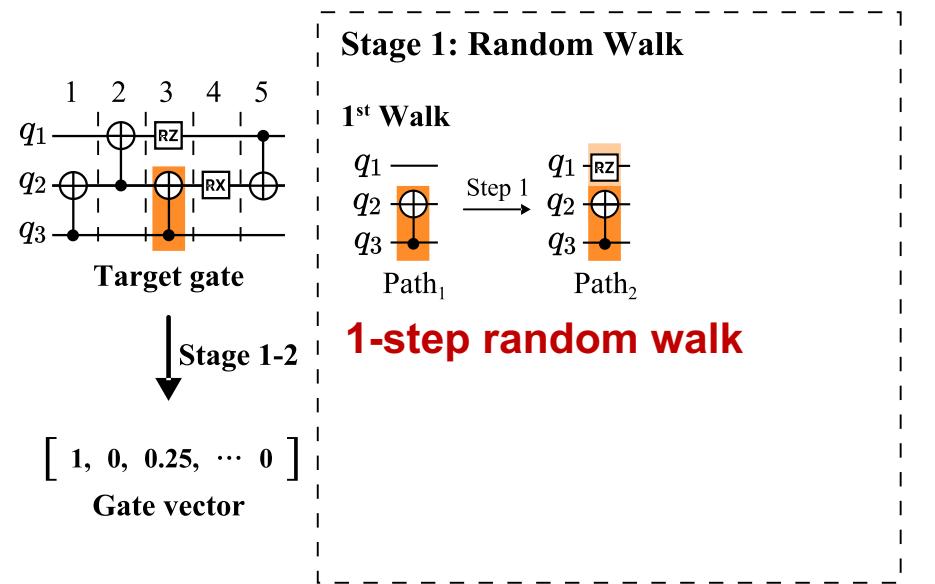
$[ 1, 0, 0.25, \dots, 0 ]$   
Gate vector

# Upstream Model: Circuit Feature Extraction



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## Two-step vectorization flow



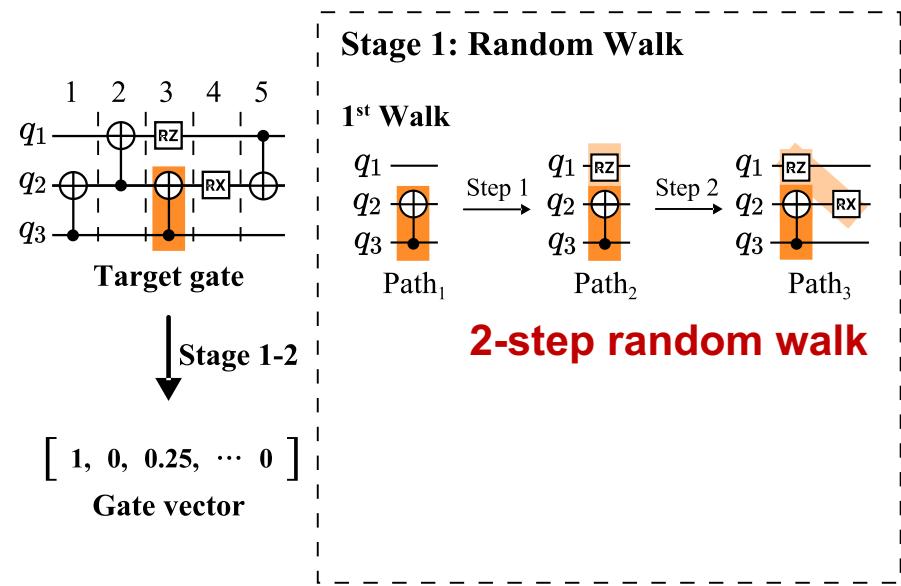
**Step 1: Extract features as paths.**

# Upstream Model: Circuit Feature Extraction



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## Two-step vectorization flow

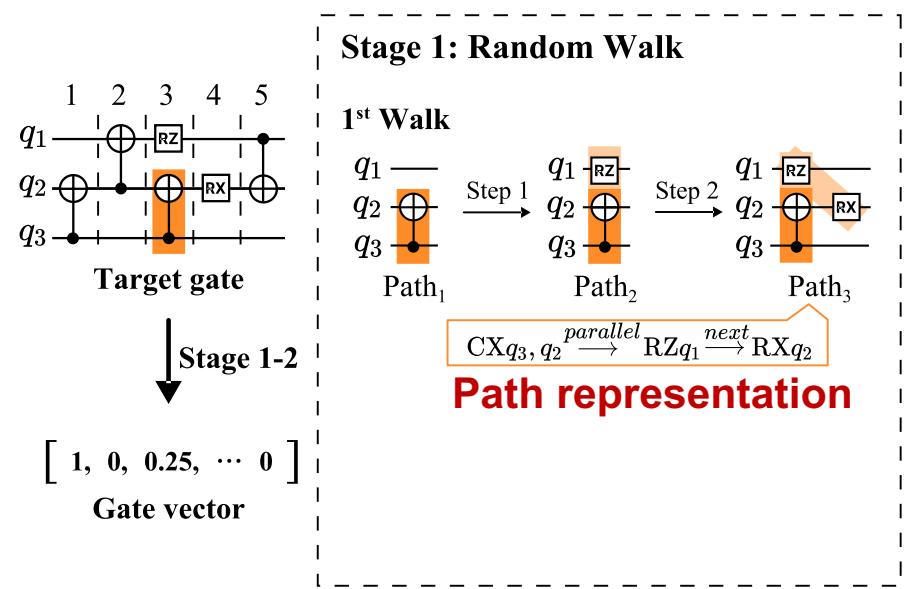


**Step 1: Extract features as paths.**

# Upstream Model: Circuit Feature Extraction



## Two-step vectorization flow

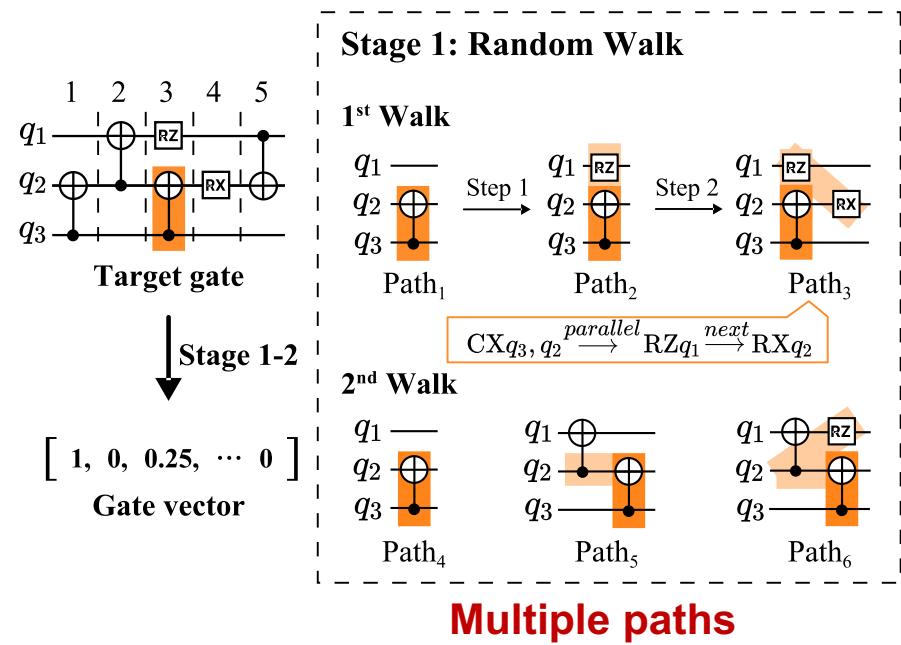


**Step 1: Extract features as paths.**

# Upstream Model: Circuit Feature Extraction



## Two-step vectorization flow

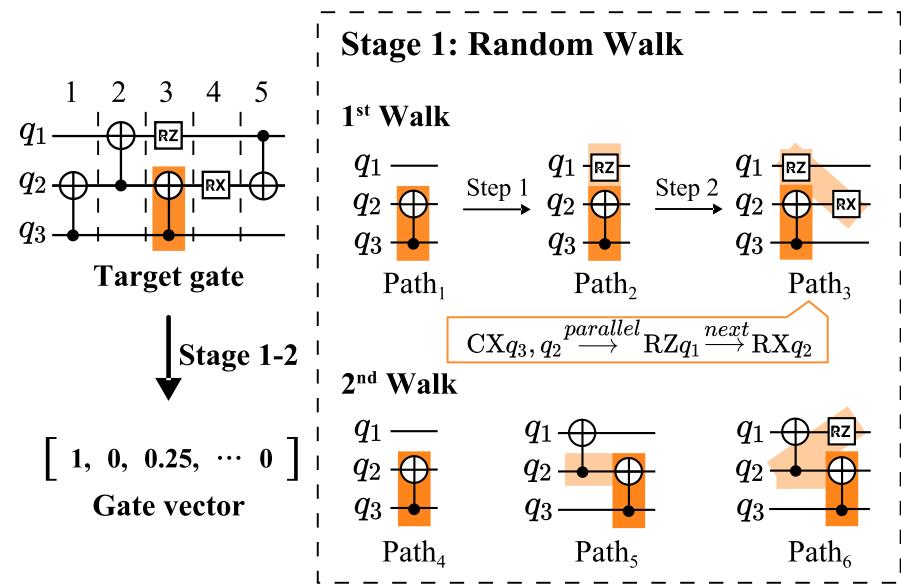


**Step 1: Extract features as paths.**

# Upstream Model: Circuit Feature Extraction



## Two-step vectorization flow

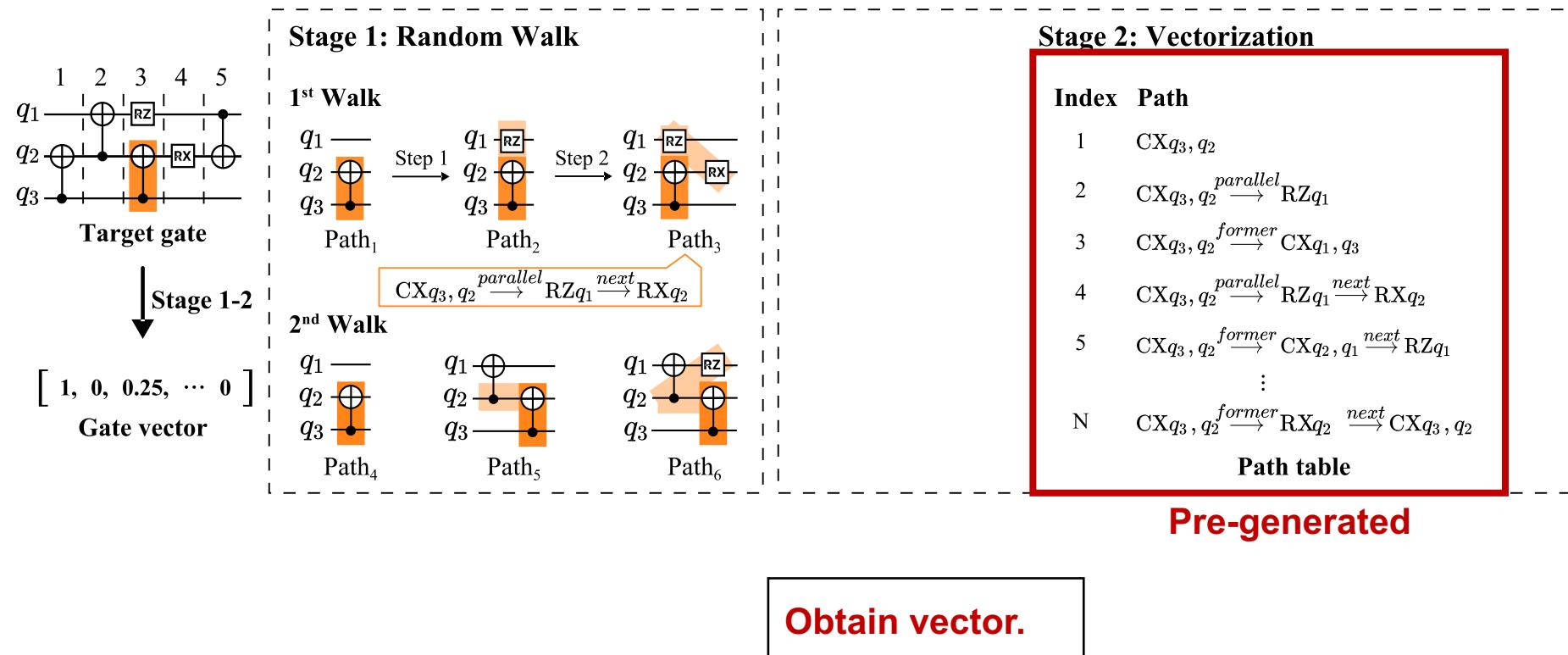


Obtain vector.

# Upstream Model: Circuit Feature Extraction



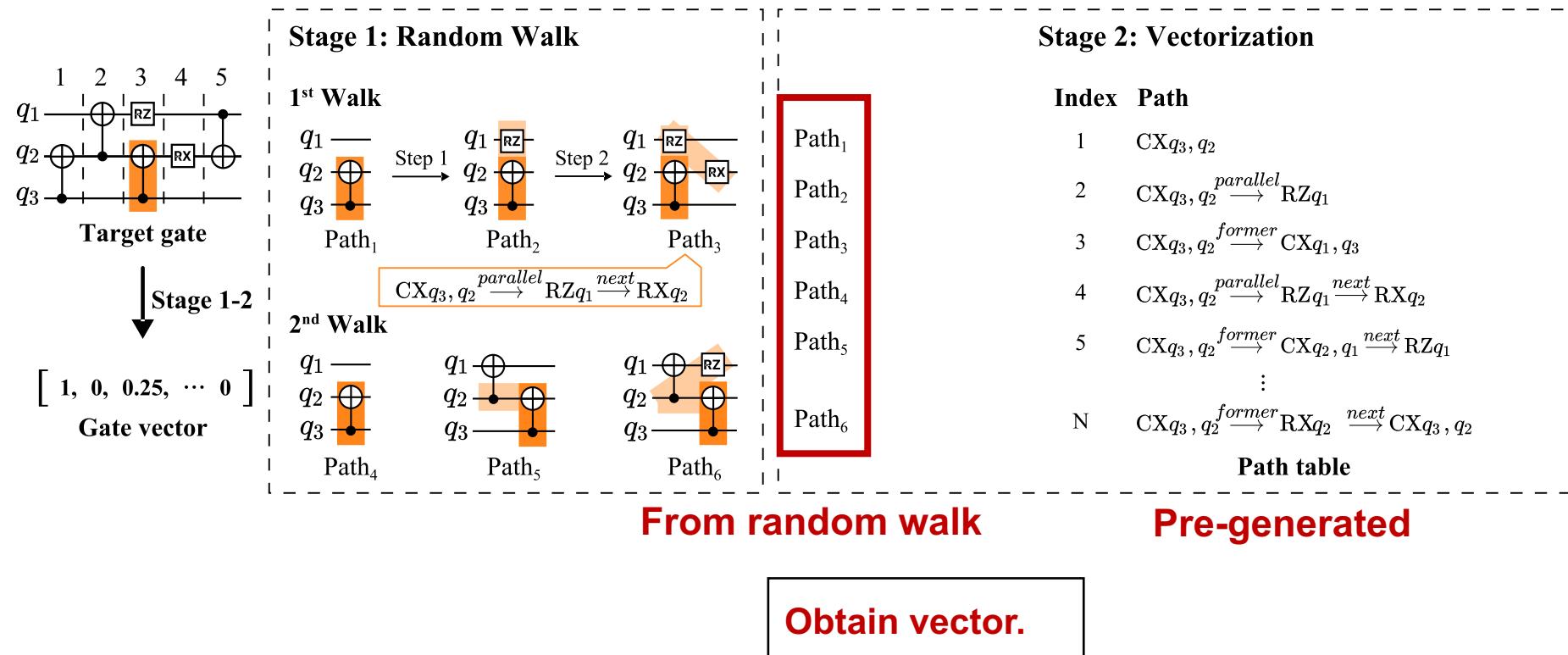
## Two-step vectorization flow



# Upstream Model: Circuit Feature Extraction



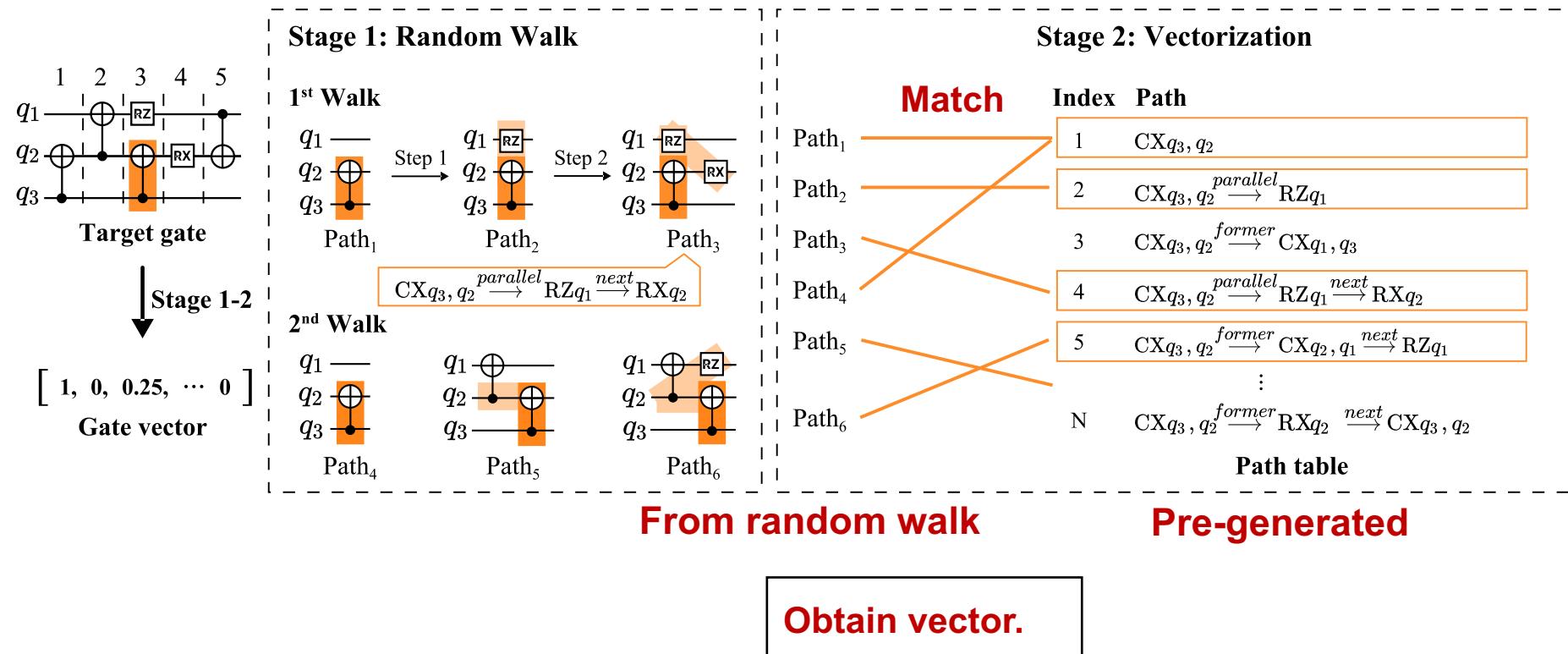
## Two-step vectorization flow



# Upstream Model: Circuit Feature Extraction



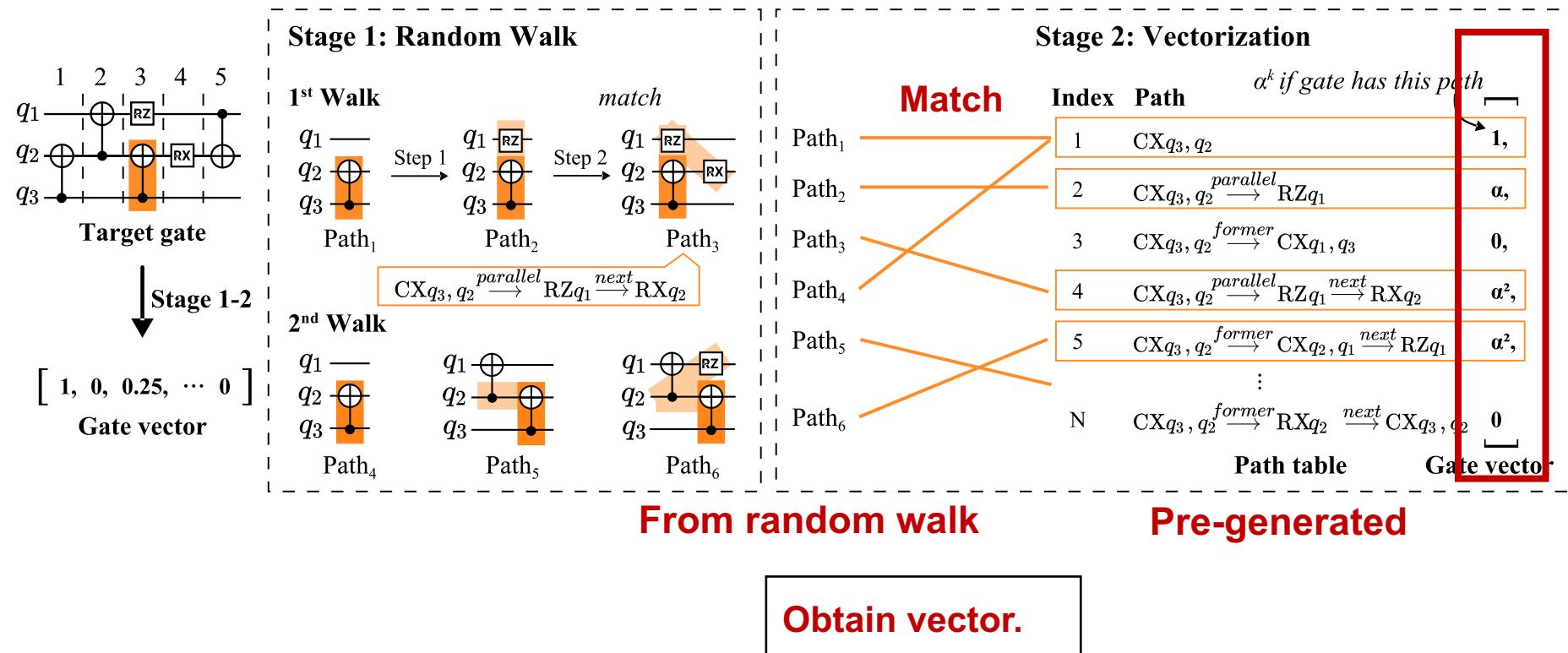
## Two-step vectorization flow



# Upstream Model: Circuit Feature Extraction



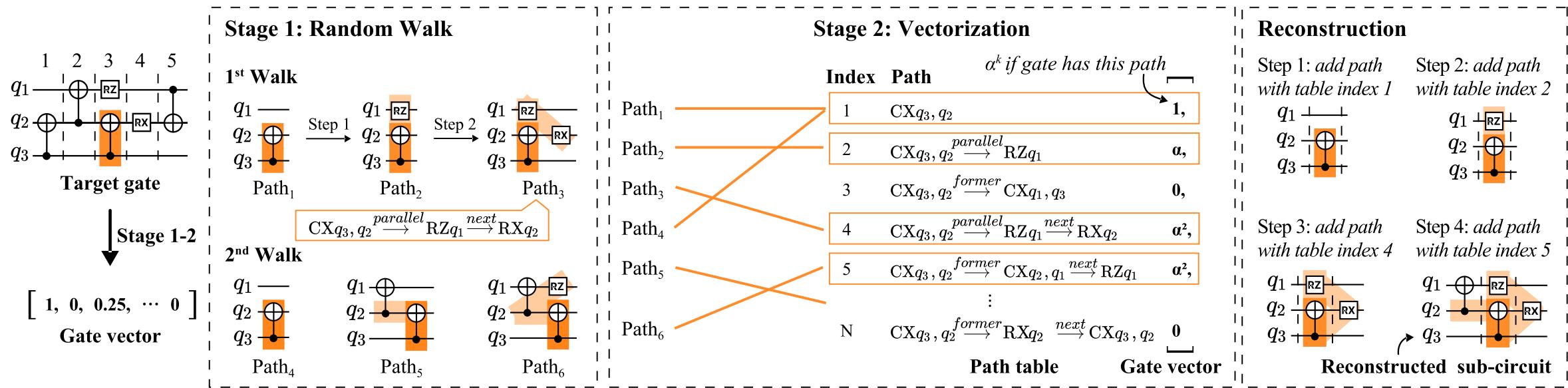
## Two-step vectorization flow



## Upstream Model: Circuit Feature Extraction



# Two-step vectorization flow



## Reconstruct circuit.

# API to Construct Upstream Model



File:

- examples/2-1.vectorization.ipynb
- <https://janusq.github.io/tutorials/Demonstrations/2-1.vectorization.ipynb>

define backend

{

```
from janusq.analysis.vectorization import RandomwalkModel, extract_device
from janusq.data_objects.backend import GridBackend, FullyConnectedBackend, LinearBackend
# define the information of the quantum device
n_qubits = 6
backend = GridBackend(2, 3)
```

generate fidelity  
dataset

{

```
# generate a dataset including varous random circuits
from janusq.data_objects.random_circuit import random_circuits, random_circuit
circuit_dataset = random_circuits(backend, n_circuits=100, n_gate_list=[30, 50, 100],
two_qubit_prob_list=[.4], reverse=True)
```

construct model  
using random  
walk

{

```
# apply random work to consturct the vectorization model with a path table
n_steps, n_walks = 1, 100
up_model = RandomwalkModel(n_steps = n_steps, n_walks = n_walks, backend = backend,
decay= 0.5, circuits = circuit_dataset )
up_model.train(circuit_dataset, multi_process=False)

print('length of the path table is', len(up_model.pathtable))
```

# Outline of Presentation

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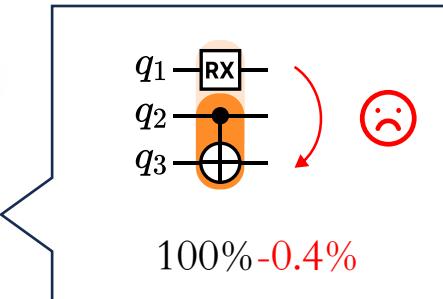
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- Downstream model 2: Unitary decomposition
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# Downstream Model 1: Circuit Fidelity Prediction



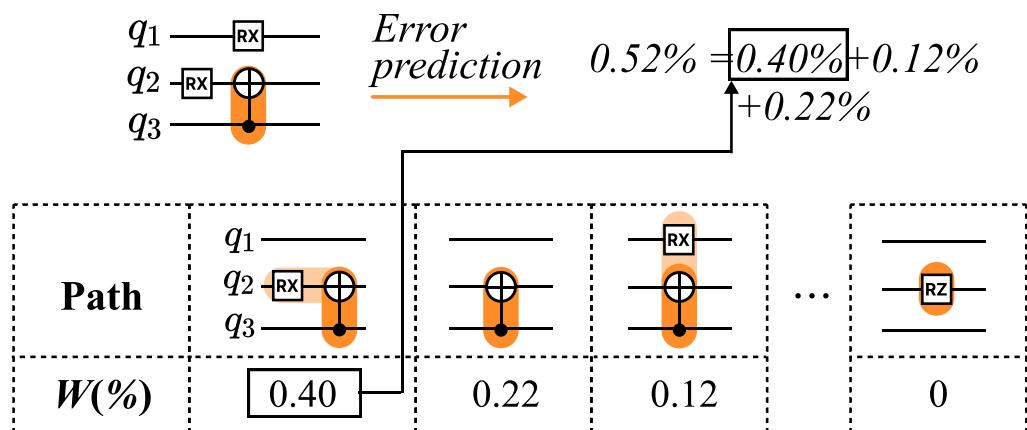
## Gate error prediction

$$E(v_i) = W^\top v_i$$



*Model dependent  
error*

$v_i$  : gate vector.  $W$  : weight vector obtained via training.



# Downstream Model 1: Circuit Fidelity Prediction



## Gate error prediction

$$E(v_i) = W^\top v_i$$

$v_i$  : gate vector.  $W$  : weight vector obtained via training.

## Circuit fidelity prediction

$$F_{\text{circuit}} = \prod_{g_i \in G} (1 - E(v_i)) \prod_{q \in Q} MF_q$$

$G$  : gate set,  $Q$  : qubit set,  $MF_q$  : measurement fidelity

***The probability that all gates are correct.***

# Downstream Model 1: Circuit Fidelity Prediction



## Gate error prediction

$$E(v_i) = W^\top v_i$$

$v_i$  : gate vector.  $W$  : weight vector obtained via training.

## Circuit fidelity prediction

$$F_{\text{circuit}} = \prod_{g_i \in G} (1 - E(v_i)) \prod_{q \in Q} MF_q$$

$G$  : gate set,  $Q$  : qubit set,  $MF_q$  : measurement fidelity

## Training process of weight vector $W$ :

Obtain fidelity dataset ( $circuit, F_{\text{ground-truth}} \dots, F_{\text{ground-truth}}$ ) : ground-truth circuit fidelity on the target quantum device.

$$\min_W |F_{\text{circuit}} - F_{\text{ground-truth}}|$$

**Minimize the distance between the prediction and ground-truth fidelity.**

# API to Construct Fidelity Prediction Model



File:

- examples/2-2.fidelity\_prediction\_simulator.ipynb
- [https://janusq.github.io/tutorials/Demonstrations/2-2.fidelity\\_prediction\\_simulator.ipynb](https://janusq.github.io/tutorials/Demonstrations/2-2.fidelity_prediction_simulator.ipynb)

Load the fidelity dataset

```
from janusq.dataset import real_qc_5bit
circuits, fidelities = real_qc_5bit
print(len(circuits))
```

Construct upstream model

```
from janusq.analysis.vectorization import RandomwalkModel
from janusq.data_objects.backend import FullyConnectedBackend
backend = FullyConnectedBackend(5)
up_model = RandomwalkModel(n_steps = 1, n_walks = 10, backend = backend, circuits =
circuits)
```

Construct fidelity prediction model

```
from janusq.analysis.fidelity_prediction import FidelityModel
fidelity_model = FidelityModel(up_model)
fidelity_model.train((circuits, fidelities), multi_process = False)
```

Predict the fidelity of the given circuit

```
circuit = random_circuit(backend, n_gates = 100, two_qubit_prob = 0.5 )
fidelity_model.predict_circuit_fidelity(circuit)
```

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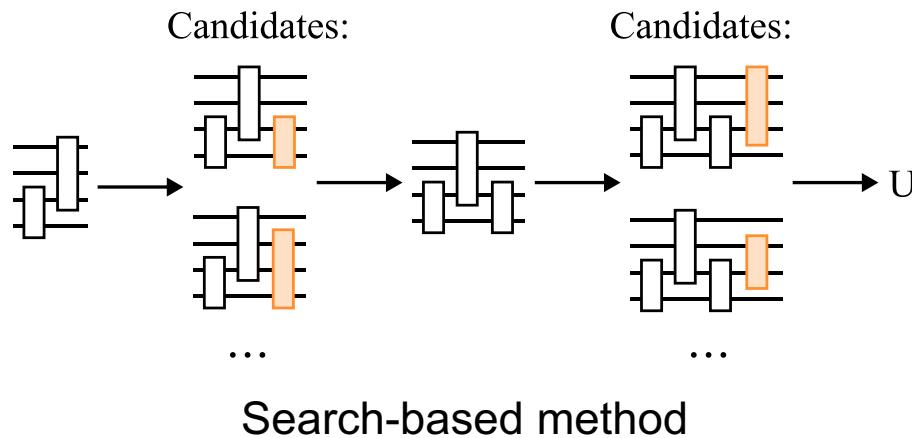


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- **Downstream model 2: Unitary decomposition**

# Downstream Model 2: Unitary Decomposition



Improve the current search-based method



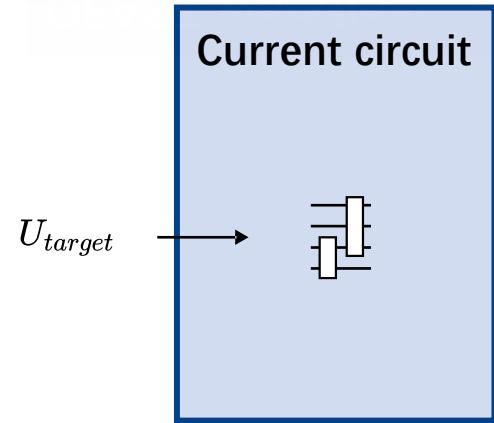
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Method	CCD [1]	QSD [2]	QFAST [3]	Squander [4]
Time	3.6 s	2.1 s	<b>511.2 h</b>	<b>426.2 h</b>
#Gate	<b>3,592</b>	<b>3,817</b>	806	887

5-qubit unitary decomposition

# Downstream Model 2: Unitary Decomposition



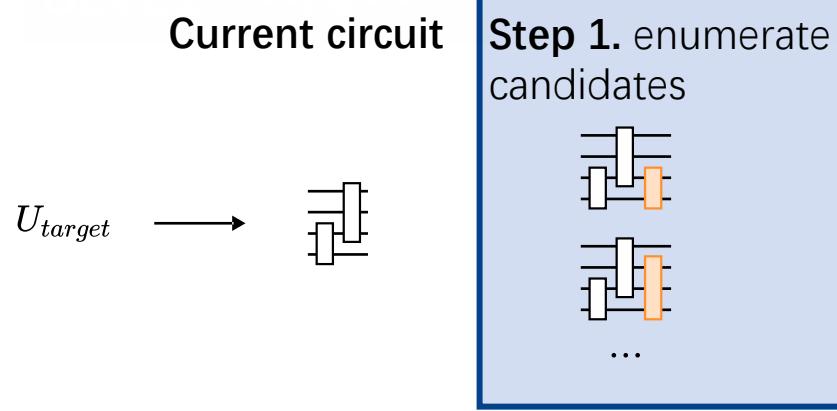
## QFAST workflow



# Downstream Model 2: Unitary Decomposition



## QFAST workflow

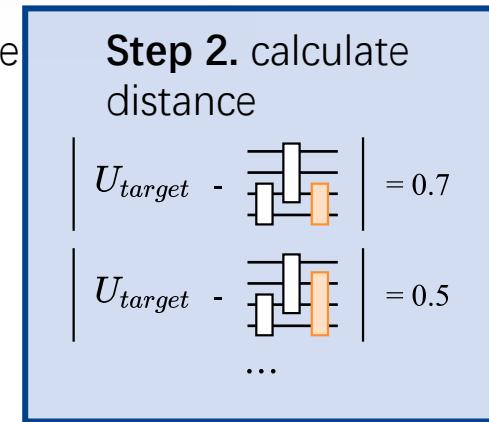
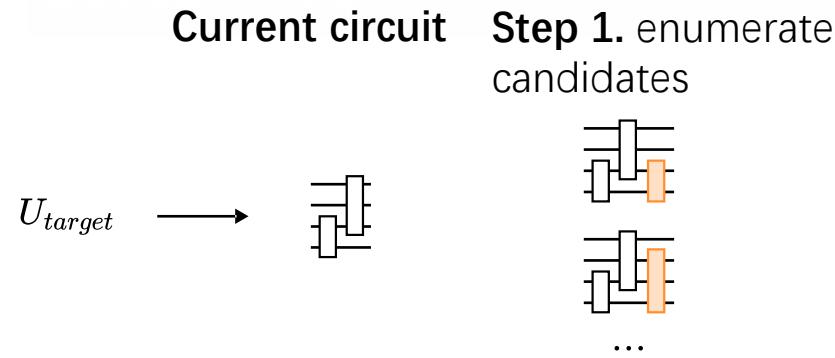


# Downstream Model 2: Unitary Decomposition



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## QFAST workflow

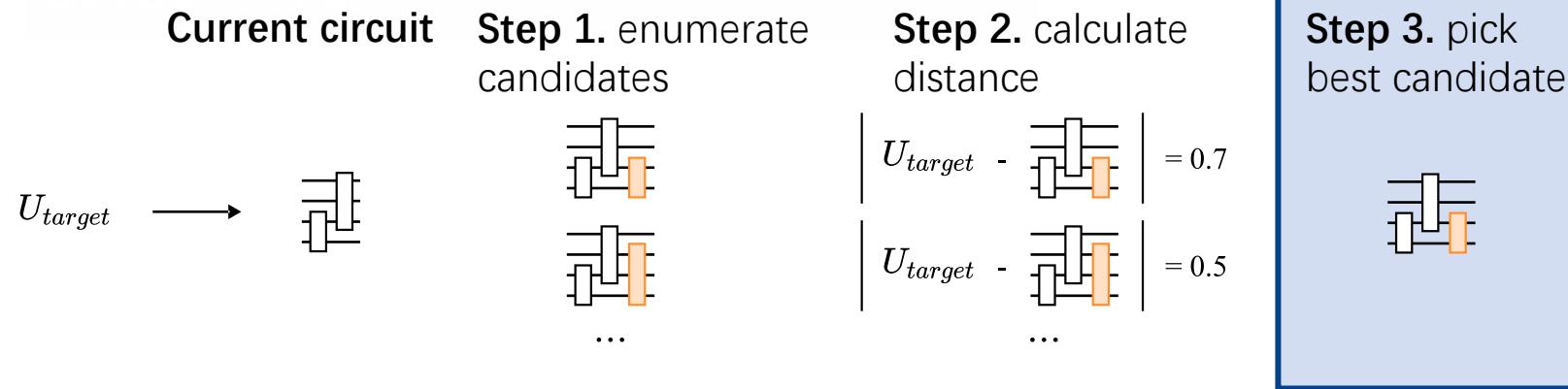


# Downstream Model 2: Unitary Decomposition



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## QFAST workflow

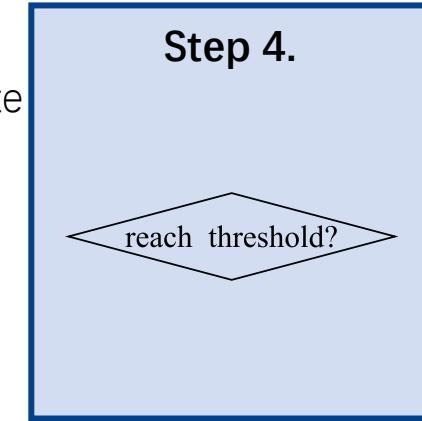
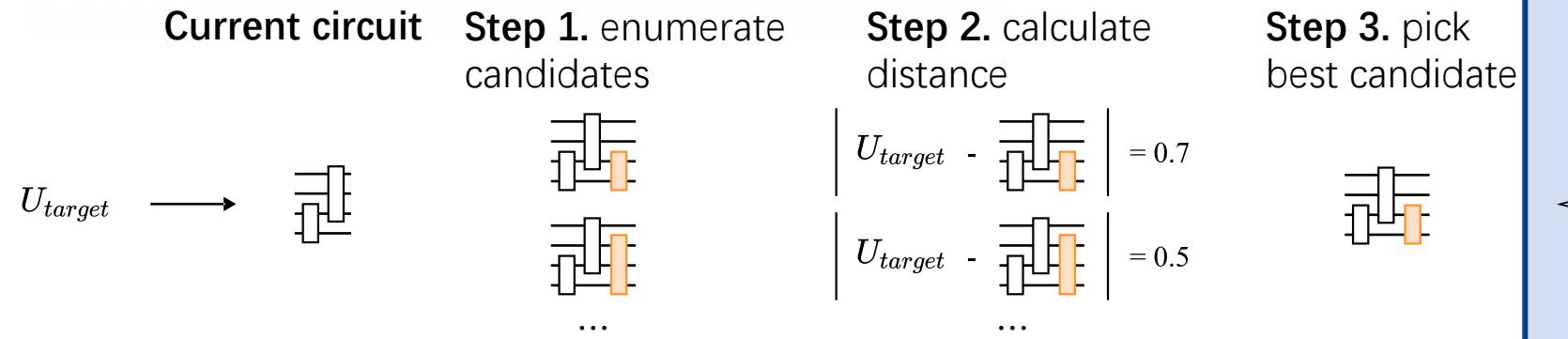


# Downstream Model 2: Unitary Decomposition



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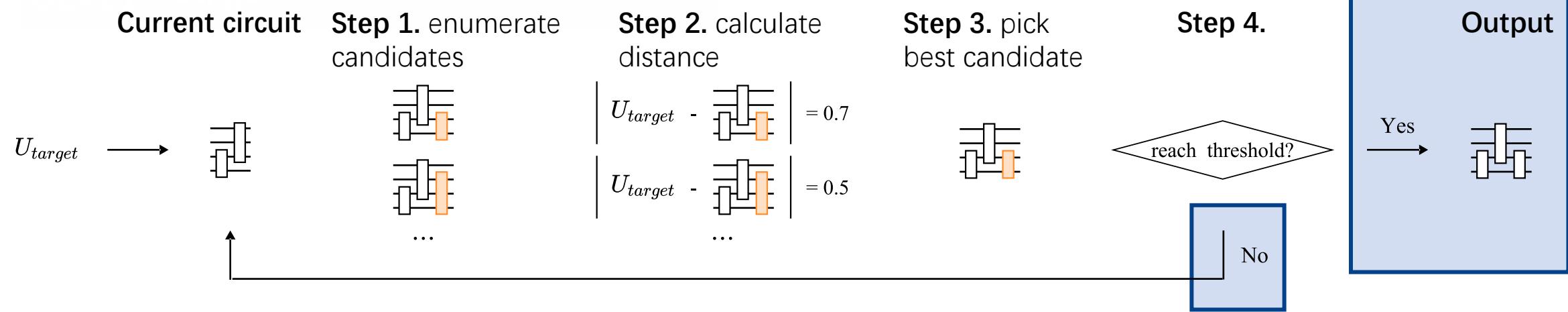
## QFAST workflow



# Downstream Model 2: Unitary Decomposition



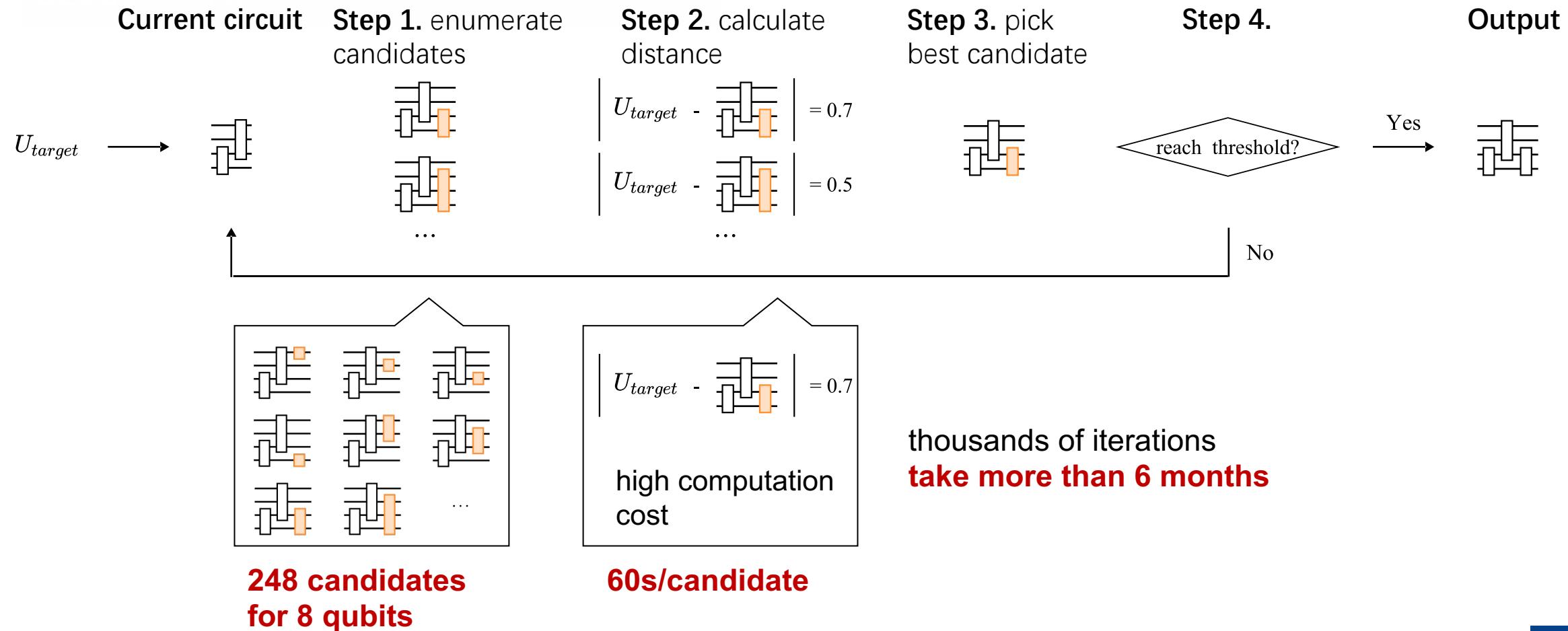
## QFAST workflow



# Downstream Model 2: Unitary Decomposition



## QFAST workflow

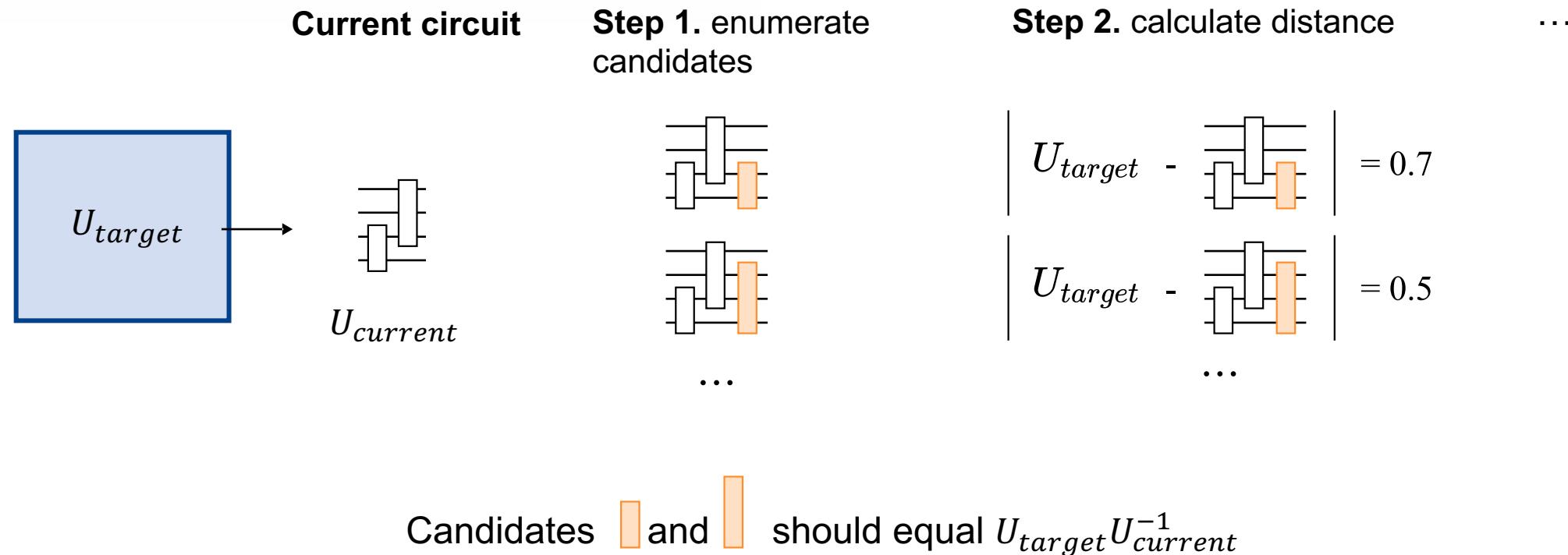


# Downstream Model 2: Unitary Decomposition



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## Janus-CT workflow

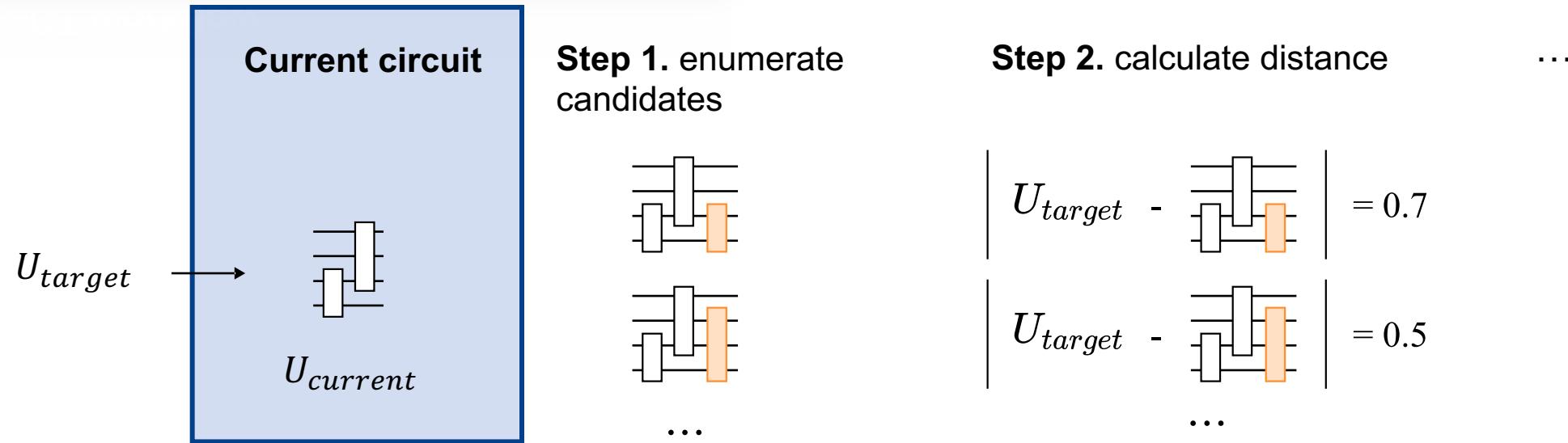


# Downstream Model 2: Unitary Decomposition



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## Janus-CT workflow



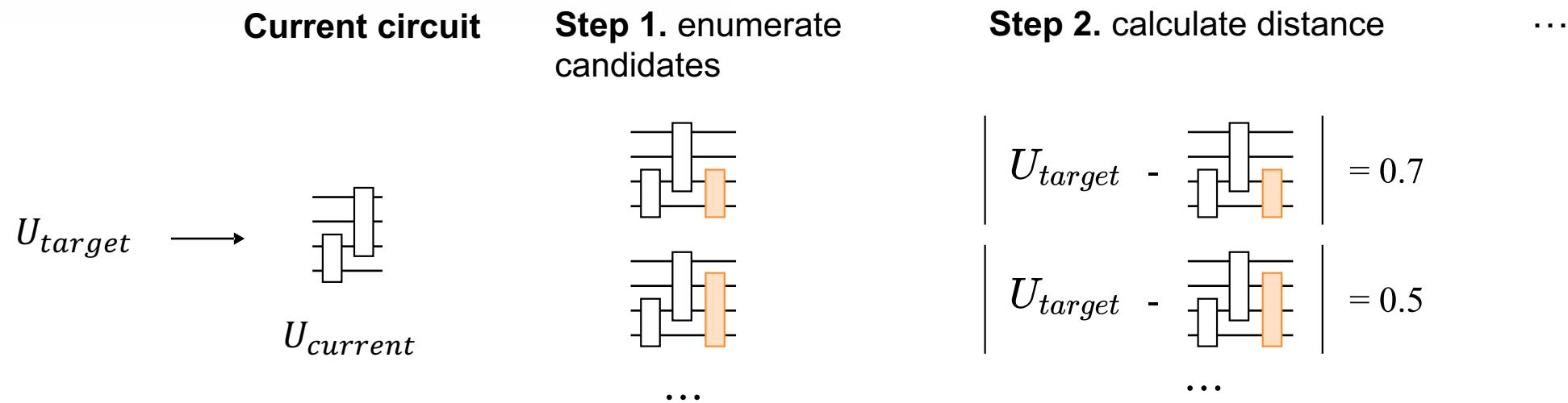
Candidates and should equal  $U_{target} U_{current}^{-1}$

# Downstream Model 2: Unitary Decomposition



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## Janus-CT workflow



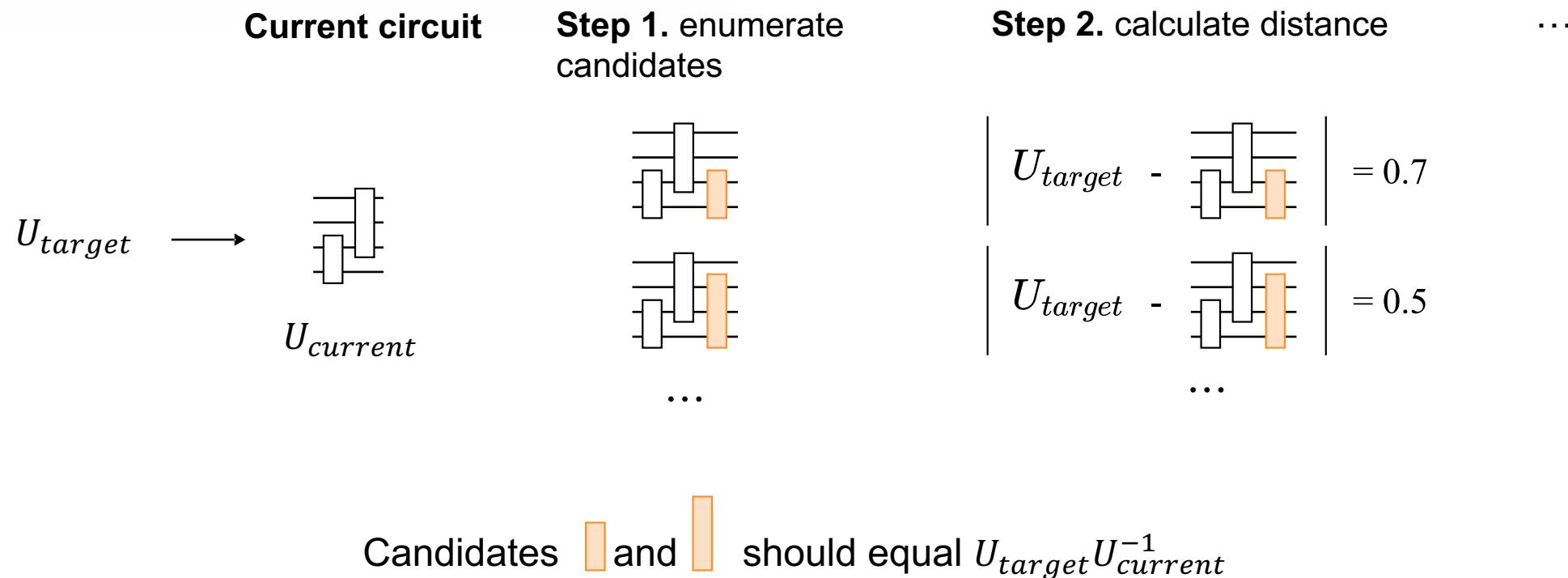
Candidates and should equal  $U_{target} U_{current}^{-1}$

# Downstream Model 2: Unitary Decomposition



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## Janus-CT workflow



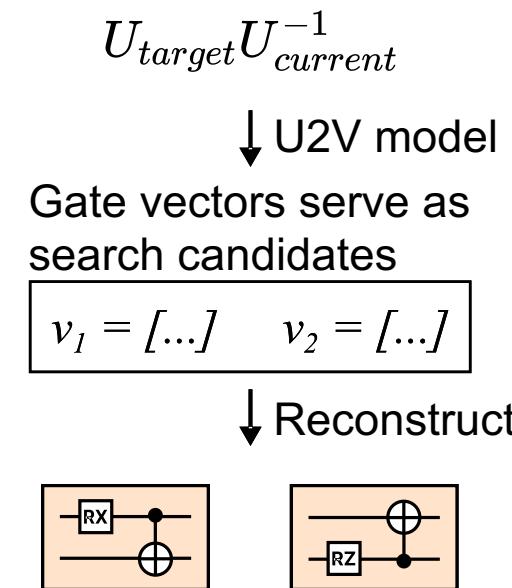
Instead of exhaustive search, Janus-CT only try the candidates that have high probability of equaling  $U_{target} U_{current}^{-1}$

# Downstream Model 2: Unitary Decomposition



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## Janus-CT workflow



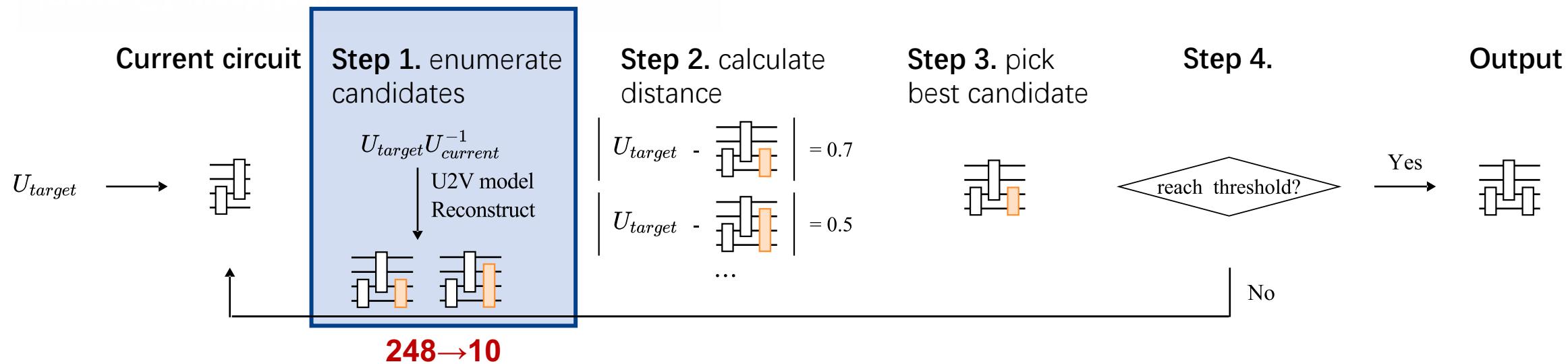
**U2V model:** a random forest model, trained by the pre-generated decomposition results.

***Use gate vectors to construct good candidates.***

# Downstream Model 2: Unitary Decomposition



## Janus-CT workflow



**Comparison to the template-based method:** template-based approach has smaller design space, as it can only selects candidates from a limited-size template library.  
1.8 $\times$  speedup and 1.6  $\times$  gate reduction compared to the template-based method.

# API to Construct Unitary Decomposition Model



File:

- examples/ 2-5.unitary\_decomposition.ipynb
- [https://janusq.github.io/tutorials/Demonstrations/2-5.unitary\\_decomposition.ipynb](https://janusq.github.io/tutorials/Demonstrations/2-5.unitary_decomposition.ipynb)

Construct the decomposition dataset

```
{ from janusq.data_objects.backend import LinearBackend
  from janusq.data_objects.random_circuit import random_circuits
  n_qubits = 4
  backend = FullyConnectedBackend(n_qubits)
  dataset = random_circuits(backend, n_circuits=50, n_gate_list=[30, 50, 100],
                            two_qubit_prob_list=[.4], reverse=True)
```

Construct unitary decomposition model

```
{ from janusq.analysis.unitary_decomposition import U2VModel
  from janusq.analysis.vectorization import RandomwalkModel
  vec_model = RandomwalkModel(2, 16, backend, directions=('parallel', 'next'))
  vec_model.train(dataset, multi_process=True, remove_redundancy=False)
  u2v_model = U2VModel(vec_model)
  data = u2v_model.construct_data(dataset, multi_process=False)
  u2v_model.train(data, n_qubits)
```

Generate a random unitary

```
{ from qiskit.quantum_info import random_unitary
  unitary = random_unitary(2**n_qubits).data
```

Apply decomposition

```
{ from janusq.analysis.unitary_decomposition import decompose
  decompose(unitary, allowed_dist = 0.2, backend = backend, u2v_model = u2v_model)
```

# Extending Janus-CT To More Downstream Tasks



File:

- examples/2-6.extend\_framework\_bug\_identification.ipynb
- [https://janusq.github.io/tutorials/Demonstrations/2-6.extend\\_framework\\_bug\\_identification](https://janusq.github.io/tutorials/Demonstrations/2-6.extend_framework_bug_identification)



```
from janusq.analysis.vectorization import RandomwalkModel

class DownstreamModel():
    def __init__(self, upstream_mdoel: RandomwalkModel):
        self.upstream_mdoel = upstream_mdoel
```

Downstream task implementation

# Extending Janus-CT To More Downstream Tasks



File:

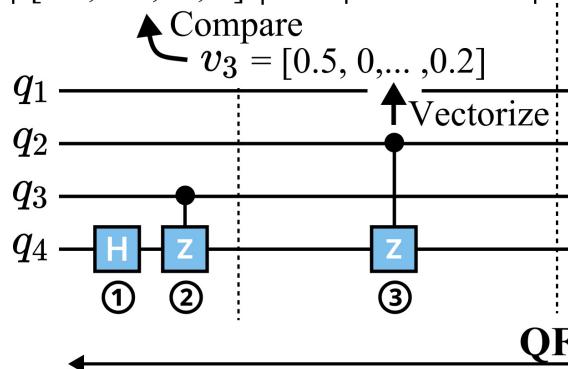
- examples/2-6.extend\_framework\_bug\_identification.ipynb
- [https://janusq.github.io/tutorials/Demonstrations/2-6.extend\\_framework\\_bug\\_identification](https://janusq.github.io/tutorials/Demonstrations/2-6.extend_framework_bug_identification)

For example: Bug Identification

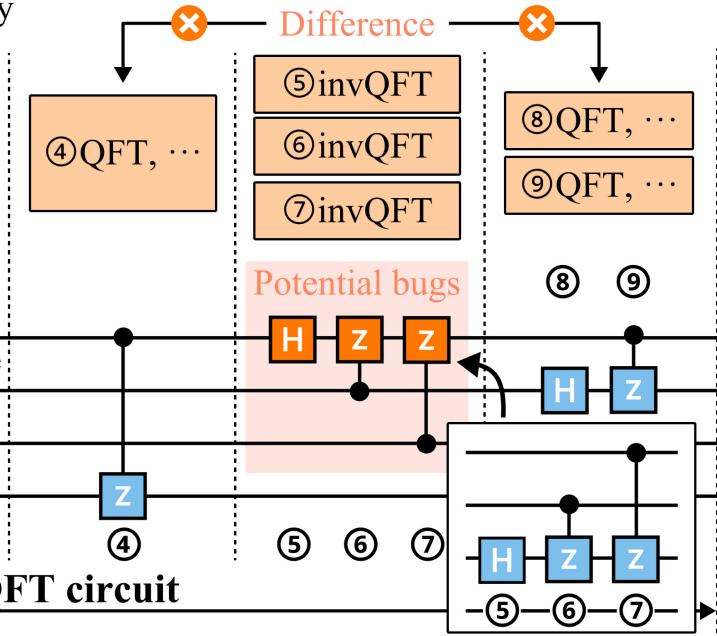
1. Identify the possible functionality

Vector list	Dist.	Functionality
[0.5, 0, ..., 0.3]	0.1	③QFT
[0.4, 0, ..., 0.2]	0.4	
[0.3, 0, ..., 0.2]	0.6	③QSVM
[0.5, 0.1, ..., 0]	0.8	③Grover

Thre.  
=0.7

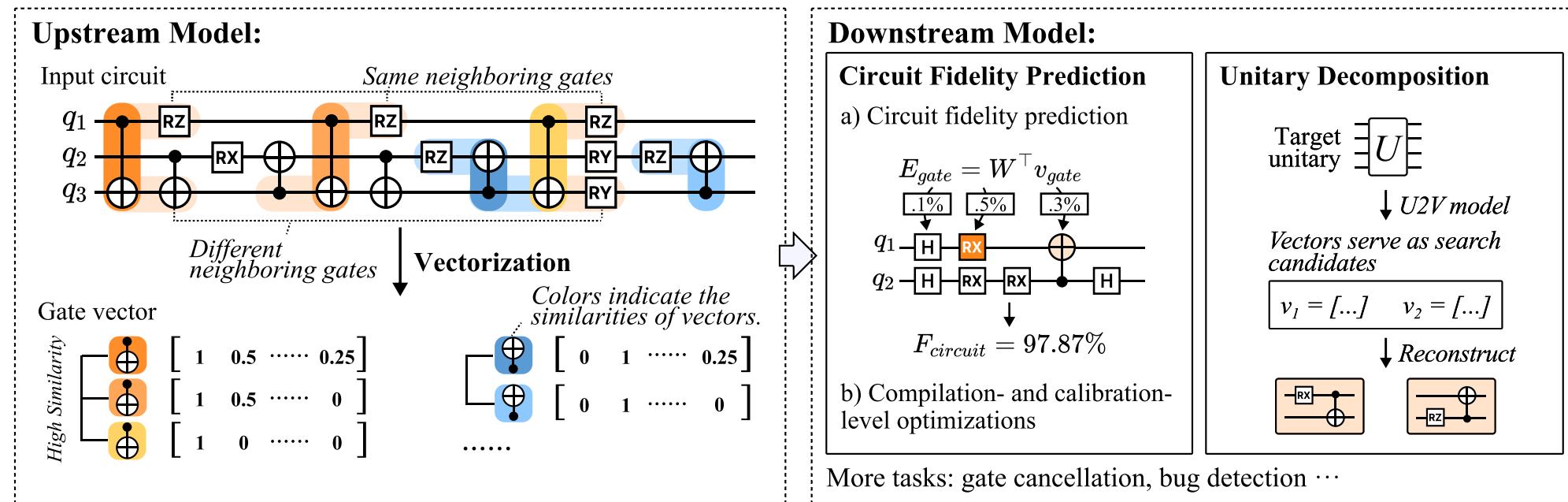


2. Identify the abnormal functionality (different from neighbors)





- Random walk-based method to extract contextual and topological circuit feature.
- Accurate circuit fidelity prediction via modeling gate interactions.
- Fast unitary decomposition via pruning candidate space.





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# Thanks for listening

**Janus-CT: A Framework for Analyzing Quantum Circuit by Extracting Contextual and Topological Features**

Siwei Tan; Congliang Lang; Liang Xiang; Shudi Wang; Xinghui Jia; Ziqi Tan; Tingting Li; Jieming Yin; Yongheng Shang,  
Andre Python, Liqiang Lu\*, Jianwei Yin\*