

# Optimizing Decision Diagrams for Measurements of Quantum Circuits

Ryosuke Matsuo<sup>1</sup>, Rudy Raymond<sup>2,3,4</sup>, Shigeru Yamashita<sup>5</sup>, Shin-ichi Minato<sup>6</sup>

1. Osaka University
2. IBM Research Tokyo
3. The University of Tokyo
4. Keio University
5. Ritsumeikan University
6. Kyoto University

# Background

Variational quantum algorithm (VQA):

- Promising application of near-term quantum computers
- Require measurement methods that efficiently extract information from quantum state

Shallow-circuit quantum measurement:

- Estimate quantum state by measuring each qubit
- Measurement based on single-qubit gate such as Pauli operations  $X, Y, Z$
- Robust to noise of quantum computers

Classical Shadow:

- Obtain the estimator by polynomial number of measurements
- There are several methods for increasing accuracy of estimator
- Decision Diagram (DD)-based method [Hillmich et al. 2021]: High accuracy

# Shallow-circuit quantum measurements

Consider  $n$ -qubit Hamiltonian  $H$  represented by following equation

$$H = \sum_{P \in \{I, X, Y, Z\}^n} \alpha_P P$$

Shallow-circuit quantum measurement problem:

Given quantum state  $\rho \in \mathbb{C}^{2^n}$ , we estimate  $\text{Tr}(H\rho)$  with small variance by **single qubit pauli-operation** and **measurement on computational basis ( $|0\rangle, |1\rangle$ )**



Optimization of **probability distribution  $\beta$**  over **measurement basis  $\mathcal{P}^n$  ( $\mathcal{P} = \{X, Y, Z\}$ )**

# Variance of estimator

$$\text{Var}(v) = \frac{1}{S} \left( \left( \sum_{P,Q} \alpha_P \alpha_Q g(P, Q, \beta) \text{Tr}(PQ\rho) \right) - (\text{Tr}(H\rho))^2 \right)$$

This is small when # of measurement basis covering  $P, Q$  is small  
 Reduces the variance

Explanation of terminology “cover”:

$\text{Cover}(P) := \{B \in \mathcal{P}^n \mid B_i = P_i \text{ whenever } P \neq I\}$

$B$  covers  $P$ .  $\implies \text{Tr}(P\rho)$  can be estimated by measurement on  $B$ .

e.g.)  $XXXX$  covers  $XIXI$

Existing decision diagram (DD)-based method [Hillmich, et al. 2021]:

- Extract measurement basis from DD representing Hamiltonian
- Demonstrate high accuracy compared with other classical shadow methods

# Extract measurement basis from DD

DD for quantum measurements:

Rooted directed acyclic multi-graph  $G = (V, E)$

Data given to edges:

- Single-qubit Pauli-operator
- Weight  $w(e) \in (0,1]$   $\longleftarrow$  Probability of selecting edge

Requirements:

- $v \in V$  has at least one out-going edge
- For out-going edge of  $v$  ( $e \in \text{out}(v)$ ),  $\sum_{e \in \text{out}(v)} w(e) = 1$

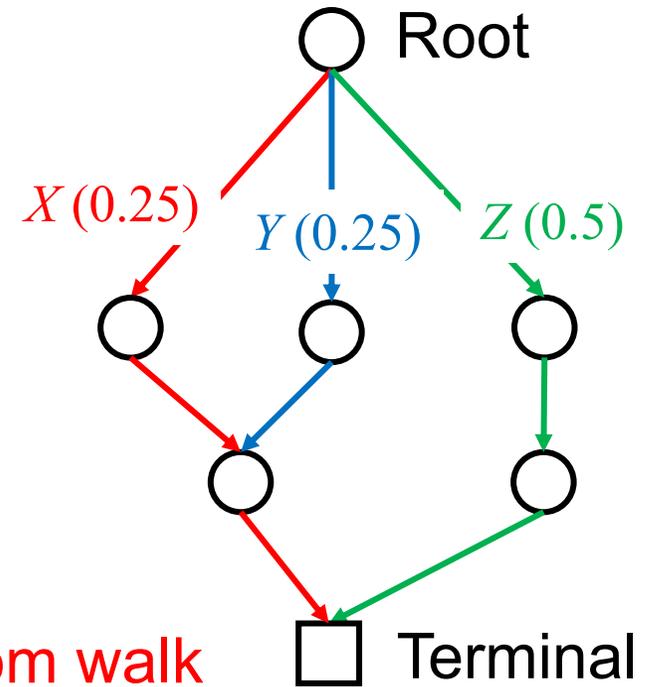
Path: Corresponds measurement basis

Estimate  $\text{Tr}(H\rho)$  using measurement basis obtain by random walk

Factor of variance: Edge weight, Shape of graph

$$H = 0.25XXX + 0.25YYX + 0.5ZZZ$$

Convert



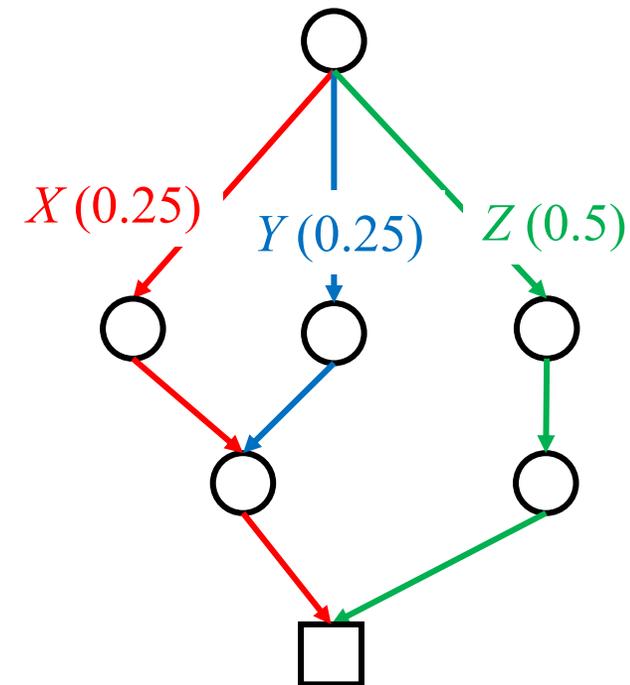
# Building method for DD

Flow of existing building method for DD

- Reduce Hamiltonian
  - Reduce # of Pauli-operators
  - Tuning probability distribution
- Initialize DD
  - Build DD from reduced Hamiltonian
- Normalize DD
  - Normalize edge weights
  - Eliminate redundant nodes

$$H = 0.25XXX + 0.25YYX + 0.5ZZZ$$

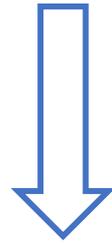
↓ Build DD



# Reduce Hamiltonian

$$H = -0.810IIII + 0.045XXXX + 0.045XXYY + 0.045YYXX + 0.045YYYY + 0.172ZIII - 0.225IZII + 0.172IIZI - 0.225IIIZ + 0.120ZZII + 0.168ZIZI + 0.166ZIIZ + 0.166IZZI + 0.174IZIZ + 0.120IIZZ$$

Reduce (Pauli grouping)



- Merge Pauli-operators
- Tuning probability distribution

Reduced Hamiltonian  $\mathcal{R}(H)$

$$= 0.045XXXX + 0.045XXYY + 0.045YYXX + 0.045YYYY + 1.714ZZZZ$$

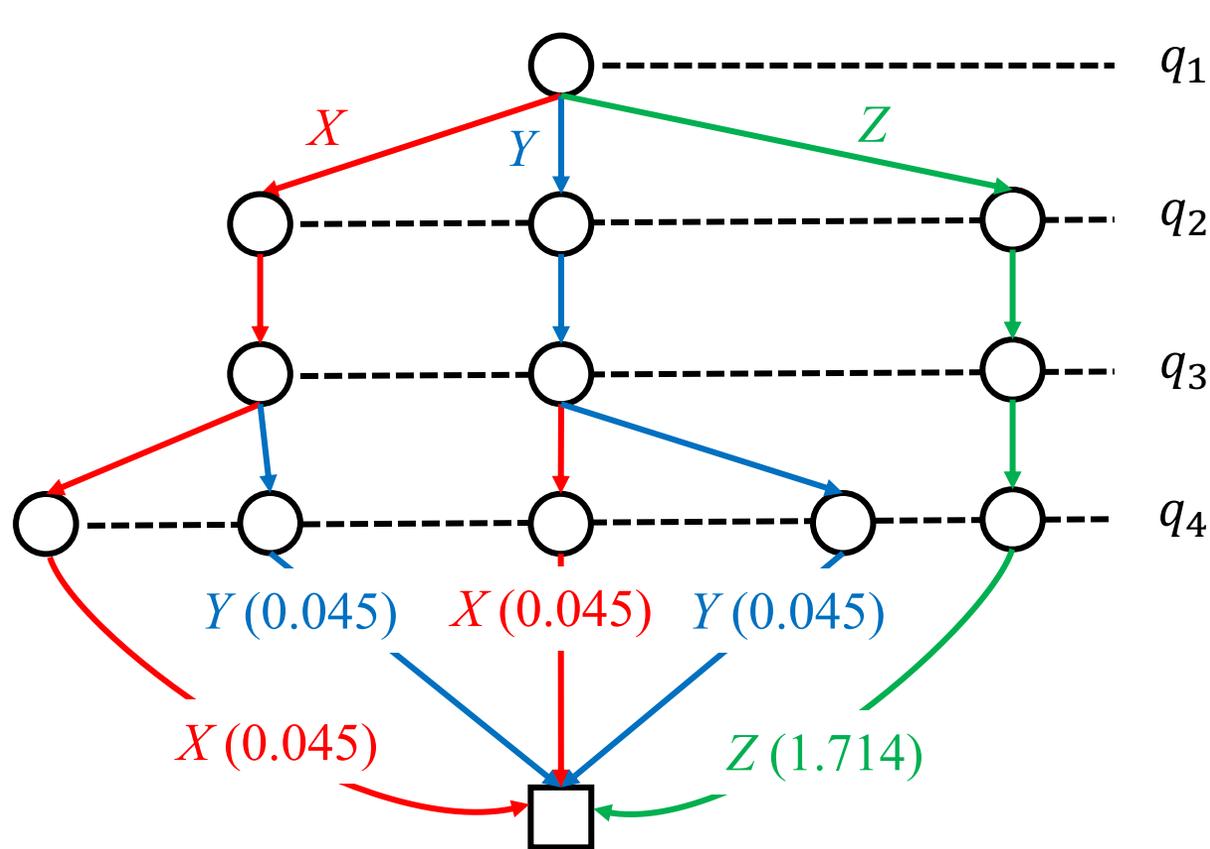
Coefficients correspond to probability distribution over measurement basis

Algorithm:

Check whether merging is possible for all combinations of terms

# Initialize DD

$$\mathcal{R}(H) = 0.045X_1X_2X_3X_4 + 0.045X_1X_2Y_3Y_4 + 0.045Y_1Y_2X_3X_4 + 0.045Y_1Y_2Y_3Y_4 + 1.714Z_1Z_2Z_3Z_4$$



Qubit order:

Order in which bits are referenced

- Create path for each pauli-operator
- Give weight to bottom edge  
(Weight corresponds to coefficient)

Next, normalize this DD

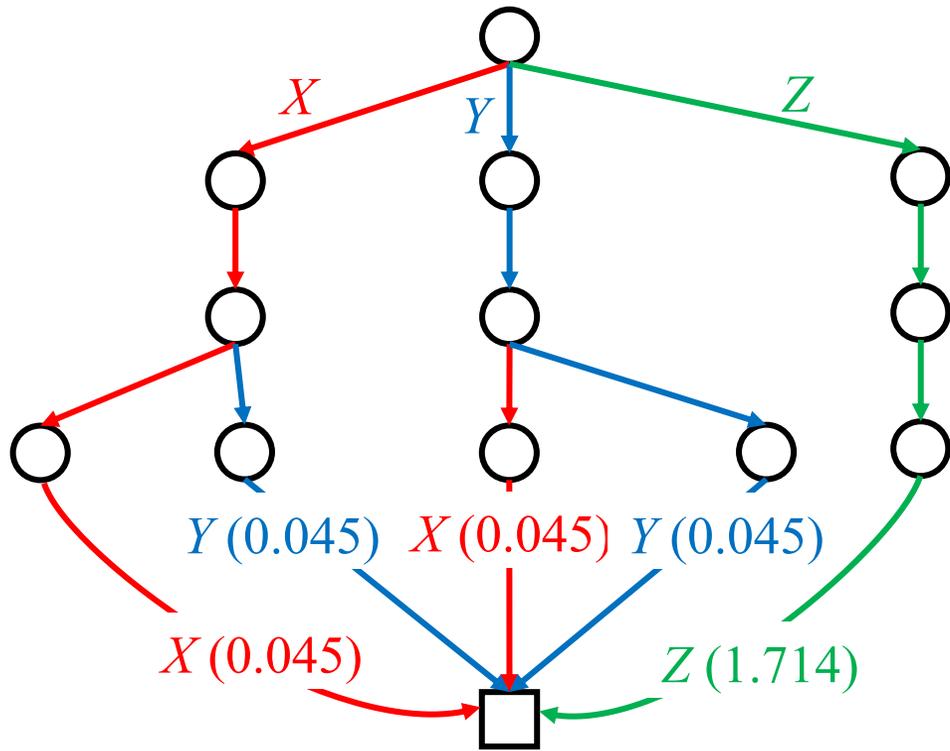
- Normalize edge weights
- Merge some nodes

# Normalize edge weight

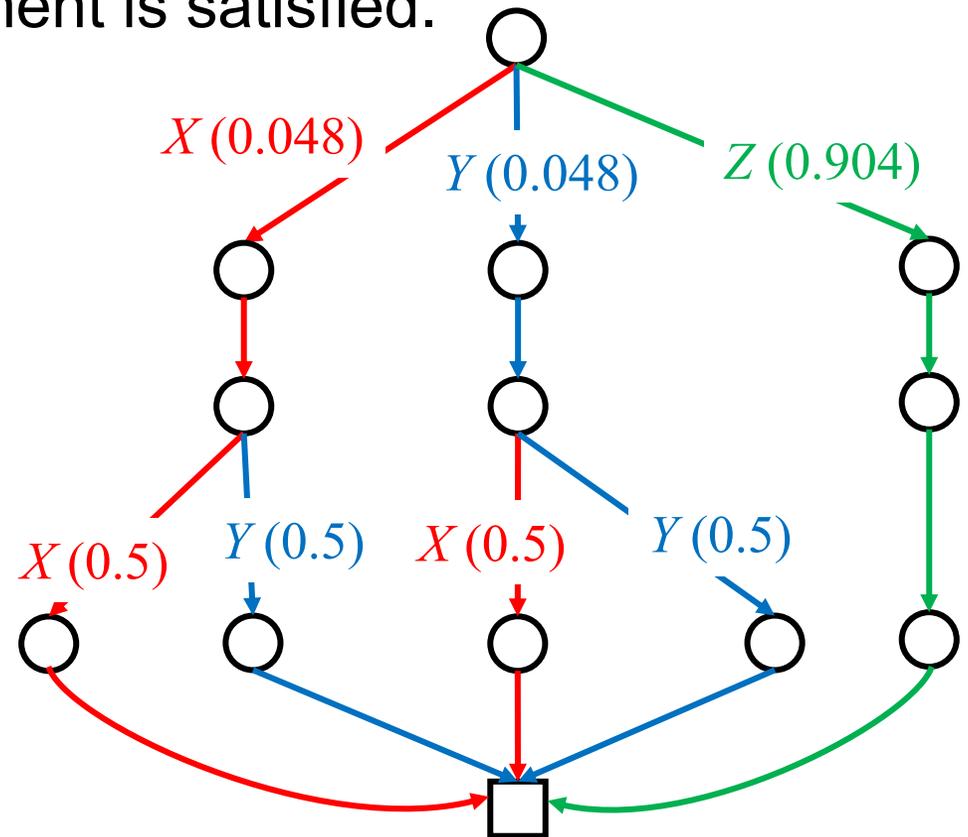
Requirement for edge weight:

For out-going edge of  $v$  ( $e \in \text{out}(v)$ ),  $\sum_{e \in \text{out}(v)} w(e) = 1$

While propagating weight to root node,  
normalize edge weight so that above requirement is satisfied.



Initial DD



Normalized weight

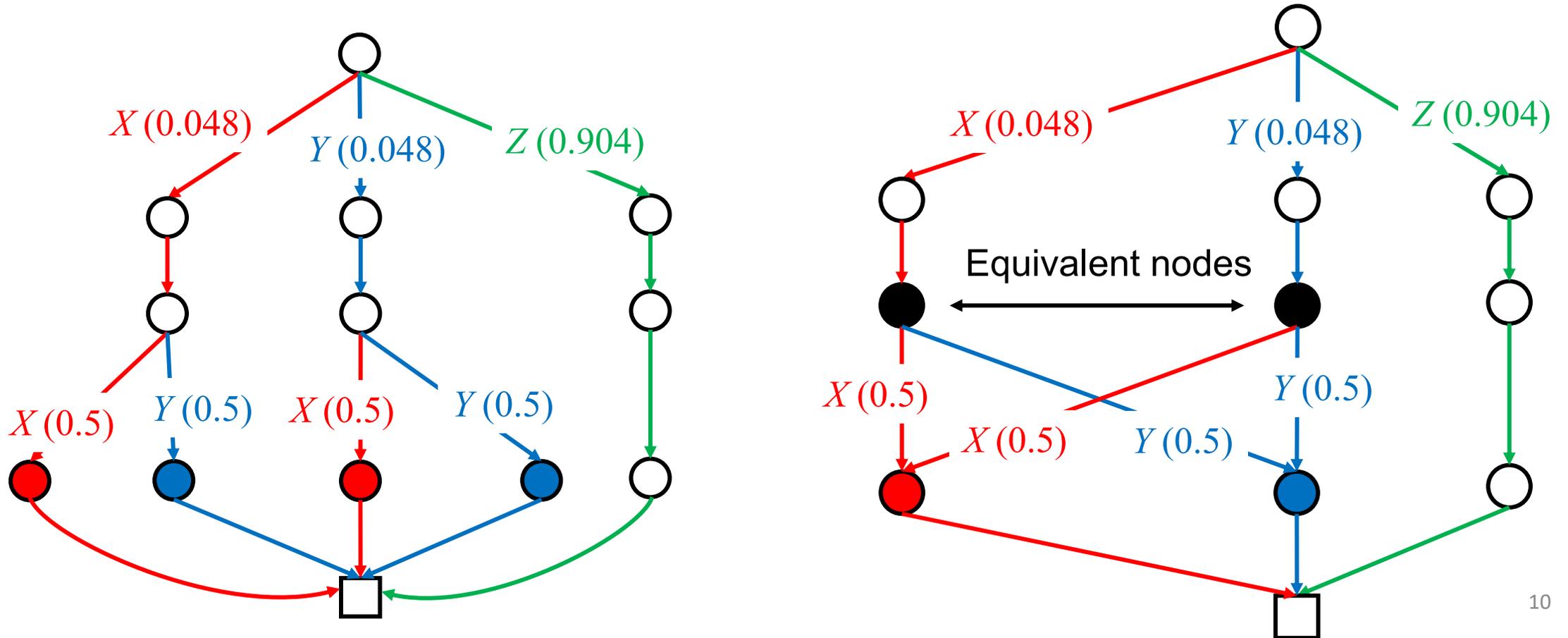
# Merge equivalent node

Merge node  $v, u$ , when the two nodes satisfy following conditions.

(1)  $V_c$  (child-nodes of node  $v$ ) =  $V_c'$  (child-nodes of node  $u$ )

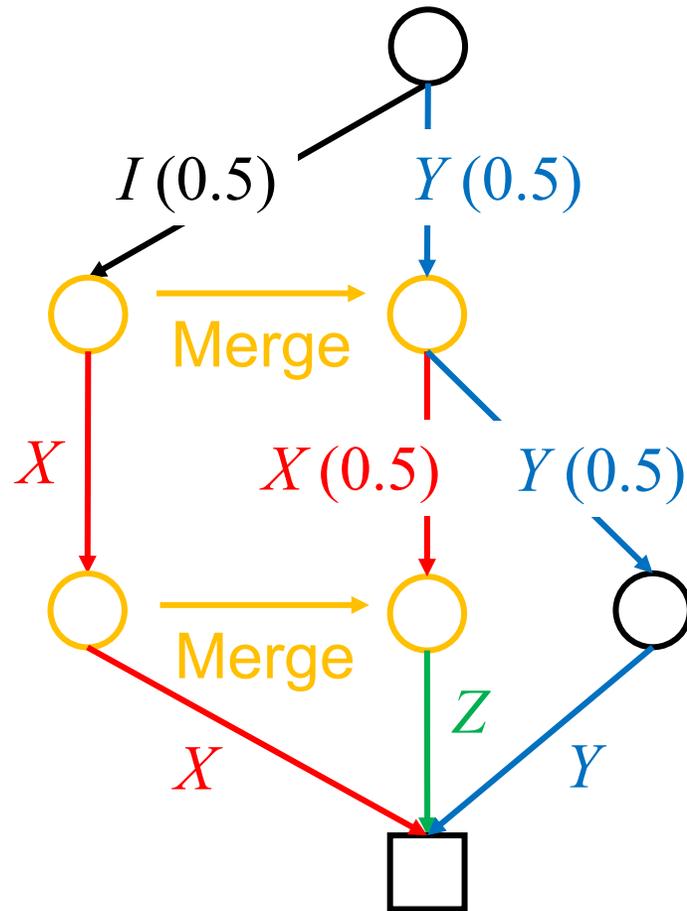
(2) For any node  $a \in V_c$ ,

**Pauli-operation** and **weight** of edge  $(v, a)$  and edge  $(u, a)$  are equivalent.



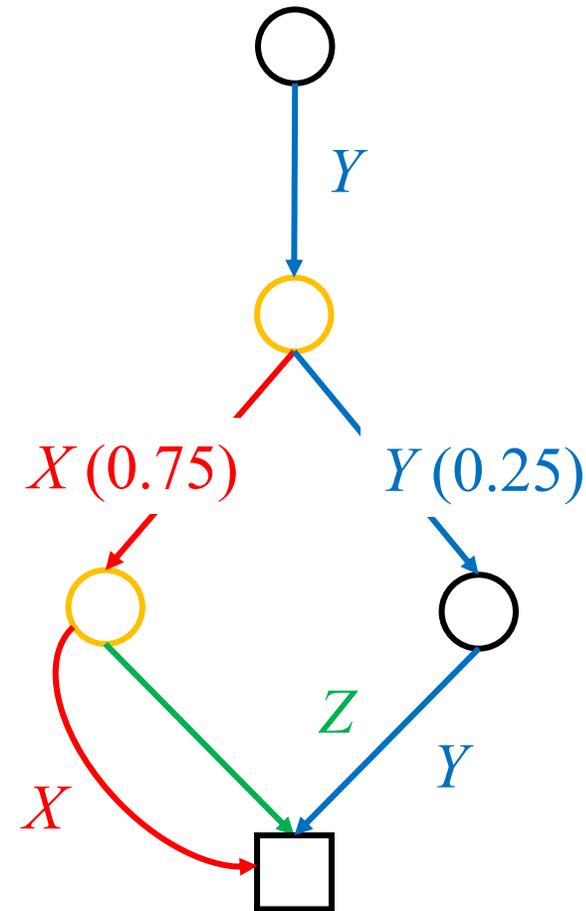
# Node reduction by merging I-edges

When graph satisfies some conditions, following operation can be applied



# of nodes = 7

Convert



# of nodes = 5

# Our contribution

Issue of DD:

Size of Hamiltonian: Large  $\Rightarrow$  Size of DD: Large

 Can not be applied to large-size Hamiltonian

Our contribution:

- Propose a method for reducing size of DD
- Our method does not compromise accuracy of estimator

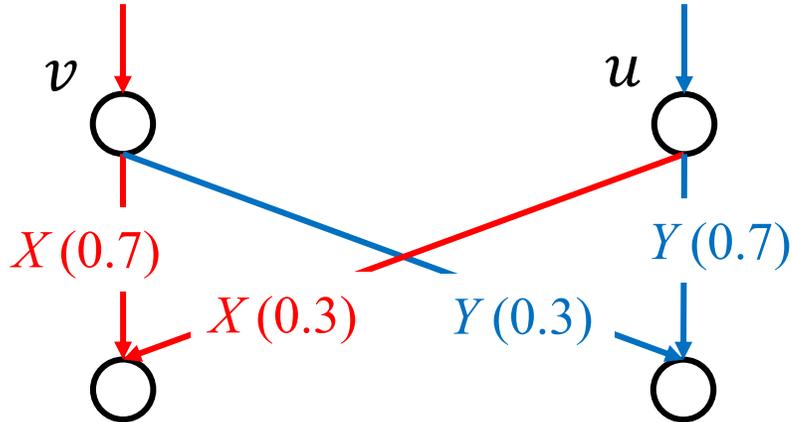
# Relaxing condition of equivalent nodes

Merge nodes  $v, u$ , when the two nodes satisfy the following conditions

(1)  $V_c$  (child-nodes of node  $v$ ) =  $V_c'$  (child-nodes of node  $u$ )

(2) For any node  $a \in V_c$ , **Pauli operations** of edge  $(v, a)$  and edge  $(u, a)$  are equivalent

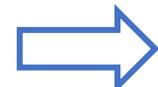
Difference from existing method: **Ignore edge weight**



Existing method:  $v, u$  are **not equivalent**

Proposed method:  $v, u$  are **equivalent**

# of nodes merged is increased

 **Reduce # of nodes in DD**

# Edge weighting scheme

$$\text{Var}(v) = \frac{1}{S} \left( \left( \sum_{P,Q} \alpha_P \alpha_Q g(P, Q, \beta) \text{Tr}(PQ\rho) \right) - (\text{Tr}(H\rho))^2 \right)$$



Small



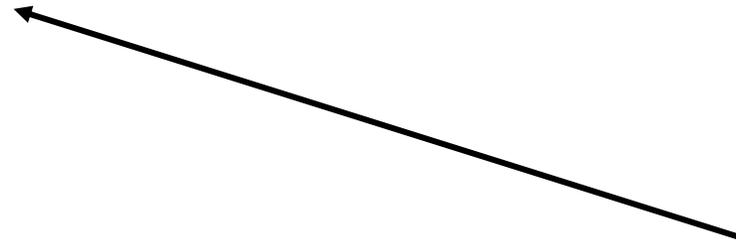
Influence of Pauli operator with large coefficient: Large



Small

# of measurement basis covering  $P, Q$ : Large

# of measurement basis covering  $P, Q$  with large coefficient: Large



Edge weighting scheme:

Give a large weight to edges in path corresponding to such measurement basis

# Qubit ordering

Reduce # of nodes by qubit ordering

DD for quantum measurements

- # of nodes: Depends on qubit order

BDD (representative DD)

- # of nodes:

Depends on variable order (corresponds to qubit order)

Naïve algorithm for finding qubit order that minimizes # of nodes:

- Examine all qubit orders

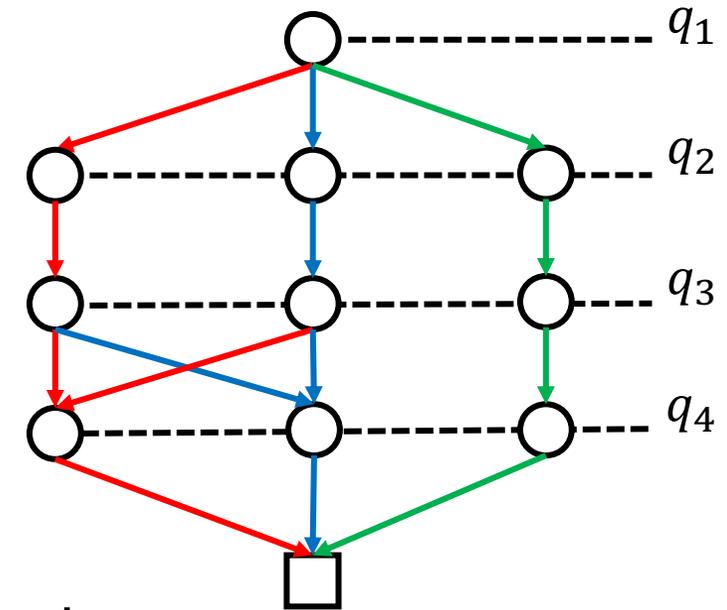
- Time complexity:  $O(n!)$

Proposed method:

- Fast algorithm based on dynamic programming

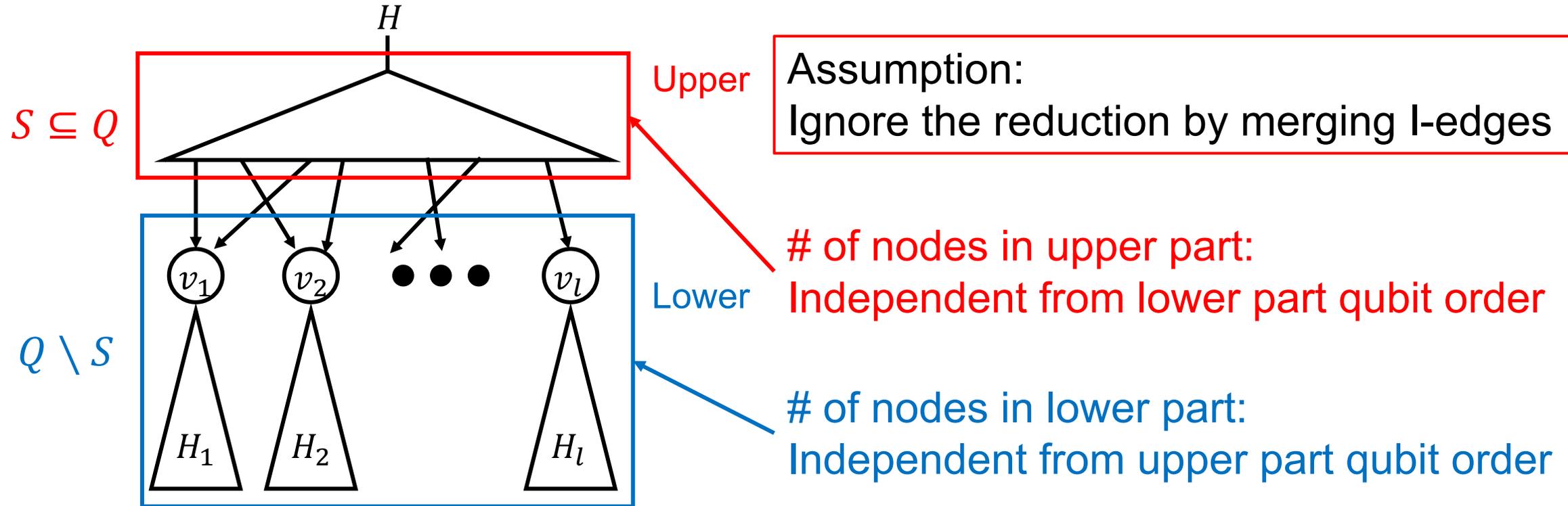
(technique utilized in BDD variable ordering)

- Obtain the optimal ordering when ignoring the reduction by merging l-edges



# Divide DD

$Q = \{q_1, q_2, \dots, q_n\}$  ( $q_i$ : Qubits in a DD)



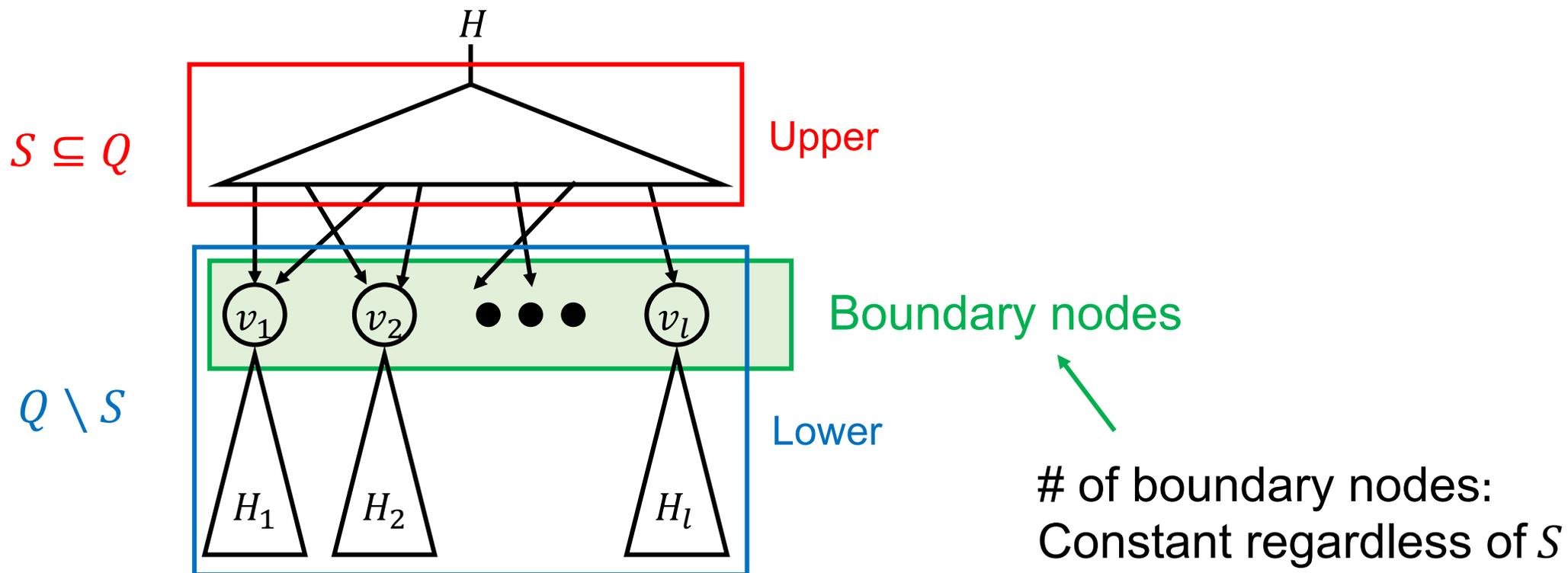
Problem of finding optimal qubit order:  
We can separate upper part and lower part

Optimal qubit order: Examine all combinations of  $S$

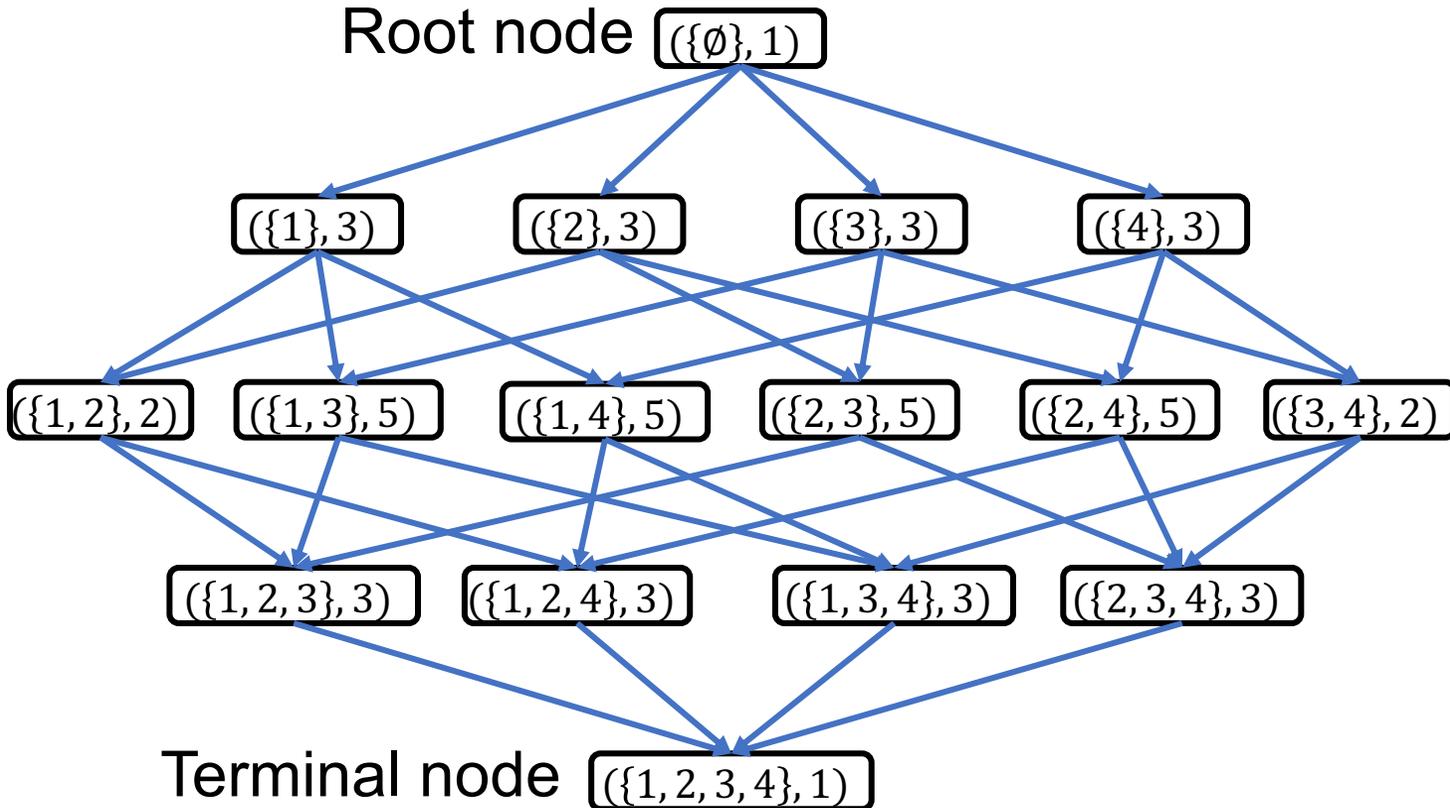
# # of boundary nodes

Information for finding optimal qubit order:

- Set of qubits  $S$
- # of boundary nodes



# Master Profile Chart(MPC)



Data of nodes: • Set of qubits  $S$   
• # of boundary nodes

Edge: Transition of  $S$

Path: Corresponds to qubit order

Path weight: # of nodes in DD  
(Node weight: # of boundary nodes)

Finding qubit order that minimize # nodes = Finding minimum weighted path

Breadth fast search can solve this problem  
Time complexity: Exponential to # of qubits

# Experimental results

Comparison of existing method with proposed method

Molecule	Encode	Terms	CS [4]	Existing DD-based [21]			Proposed		
			Vari.	Node	Edge	Vari.	Node	Edge	Vari.
LiH (12-qubit)	JW	751	266	166	271	11.3	74	137	8.28
	Parity	778	760	215	429	31.2	172	316	36.7
	BK	765	163	290	521	29.6	167	300	44.1
BeH <sub>2</sub> (14-qubit)	JW	796	1792	319	484	123	177	297	314
	Parity	831	3524	375	702	635	266	468	697
	BK	833	2273	336	545	561	226	390	3597
H <sub>2</sub> O (14-qubit)	JW	1302	4789	389	682	1444	220	397	1474
	Parity	1332	11209	540	1065	3552	229	492	5660
	BK	1333	26004	476	864	5325	265	487	9481

Proposed methods: Reduce DD size without compromising accuracy

[4] S. Aaronson, "Shadow tomography of quantum states," 2020.

[21] S. Hillmich, et. al, "Decision diagrams for quantum measurements with shallow circuits," 2021.

# Conclusion

DD based quantum measurement:

- High accuracy
- Scalability issue regarding DD size

Proposed methods:

- Relax condition of equivalent nodes
- Edge weighting scheme
- Qubit ordering optimization

Experimental results:

- Demonstrate proposed methods can reduce DD size without compromising accuracy