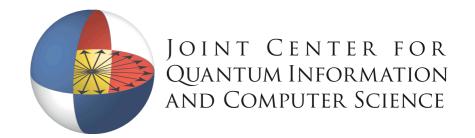
Automating NISQ Application Design w/ Meta Quantum Circuits with Constraints

Xiaodi Wu QuICS & UMD

Joint work with Haowei Deng, Yuxiang Peng, and Mike Hicks



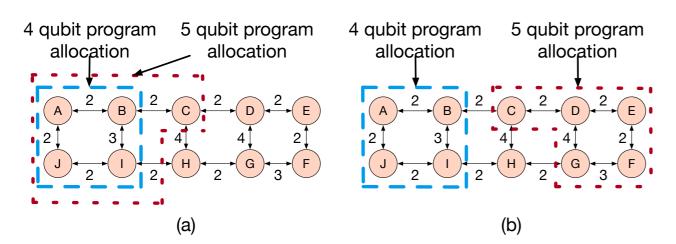


Features of NISQ Application Design

NISQ machines: very *restricted* hardware resources, where precisely controllable qubits are *expensive*, *error-prone*, *and scarce*.

NISQ application design: investigate the best balance of trade-offs among a large number of (potentially heterogeneous) factors specific to the targeted application and quantum hardware.

Multi-Programming (MICRO 2019):

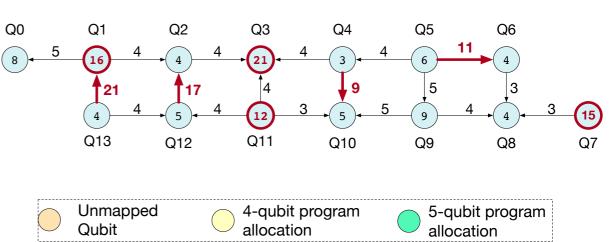


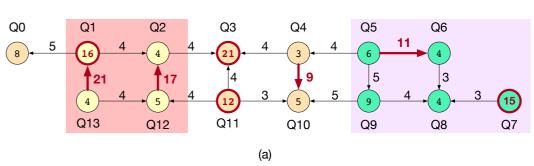
Competing Goals:

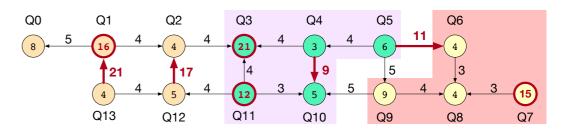
- (1) Fully leverage qubits & Shorten the total execution => Multi-Programming
- (2) High Reliability => Use the best qubits=> Sequentially Allocate Programs

Solution: A run-time trade-off between these competing goals.

IBM Q16

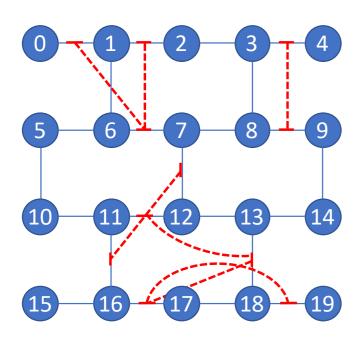






Features of NISQ Application Design

Cross-talk:



Cross-Talk: Red Pairs of gates when executed simultaneously will cause much larger errors.

IBMQ Boeblingen

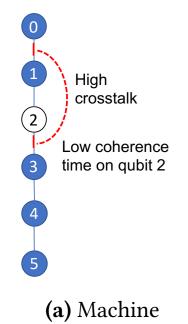
Competing Goals:

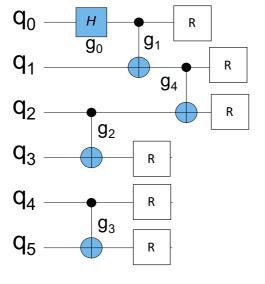
Circuit Depth (decoherence) vs Cross-Talk

Software Solutions:

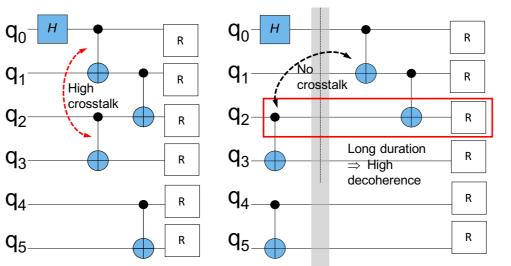
- (1) Circuit Reschedule Xtalk (ASPLOS 2020)
- (2) Frequency-Aware Compilation (MICRO 2020)

Xtalk

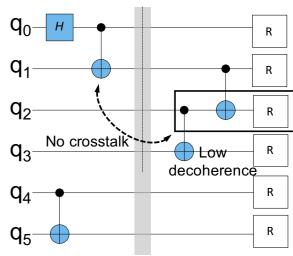








(c) Original Default Sched- **(d)** High decoherence schedule ule



(e) Desired Schedule

Automating NISQ Application Design

Current implementation of NISQ application design are CASE by CASE.



A unified and automatic framework for productivity?

Desiderata:

Succinct Expression

of different design choices

Flexible Expression

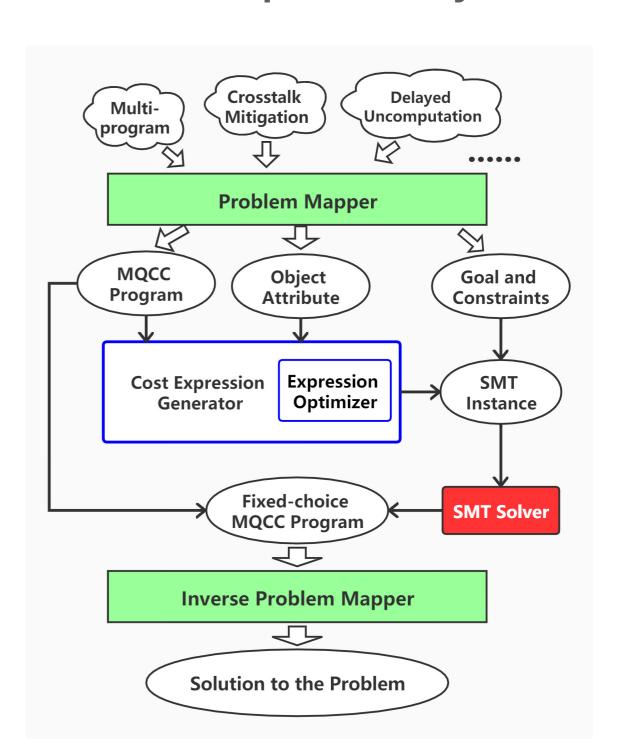
of different optimization goals

Automation of Trade-offs

of competing optimization goals

High Reusability & Productivity

of balancing different trade-offs



Desiderata:

Succinct Expression

of different design choices

MQCC with choice variables

Flexible Expression

of different optimization goals

Flexible Attributes Expression

Automation of Trade-offs

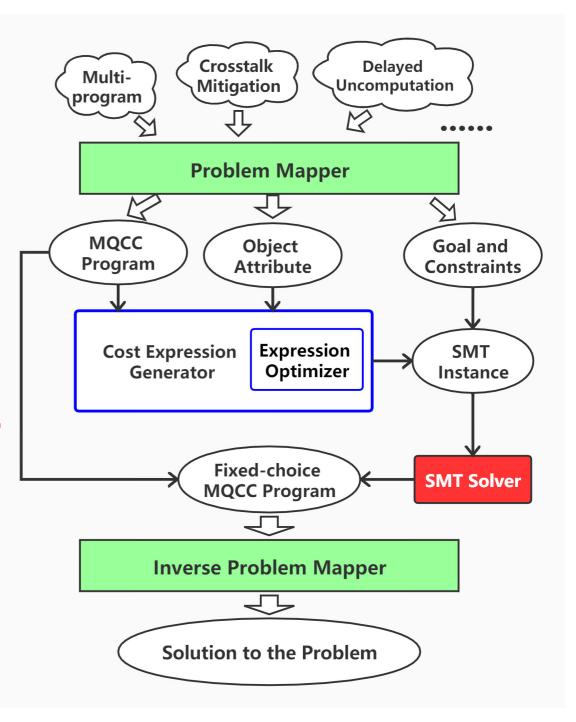
of competing optimization goals

Satisfiability Modulo Theories (SMT) Solver

High Reusability & Productivity

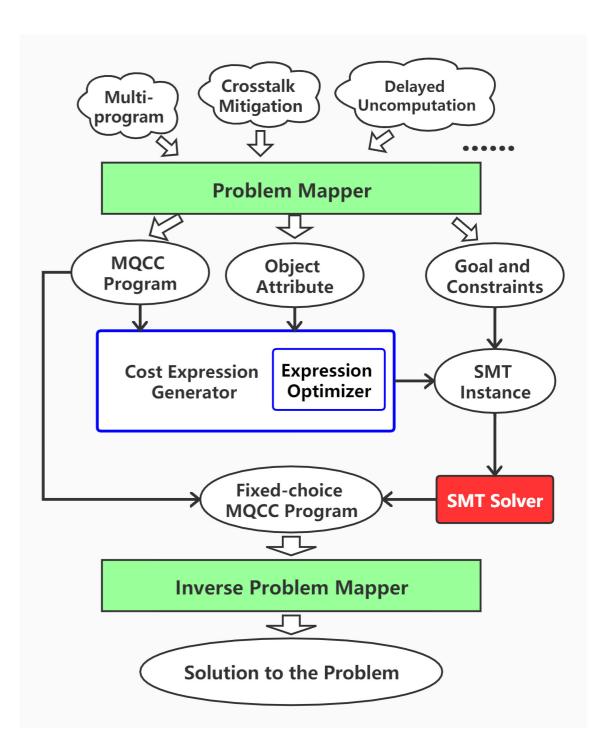
of balancing different trade-offs

A Meta-Programming Framework



```
1 \\Register and variable declarations
2 greg g[10];
3 creq r[1];
4 fcho c1 = \{0, 1\};
5 \text{ fcho } c2 = [0, 1];
6 \setminus 1 cho c = 1 - c1 * c2;
  \\Module define
9 module Bell1(q1,q2){
      h(q1);
      cnot(q1, q2);
12
13
14 module Bell2(q1, q2) {
      case (r[0]) {
           1: x(q1);
           0: pass
      } ;
      h(q1);
      cnot (q1, q2);
21
22
  \\Main part of the program
  choice (c1) {
      0: Bell1(q[1], q[2]);
      1: Bell1(q[7], q[8]);
27 };
29 h (q[0]);
30 measure (q[0], r[0]);
31 choice (c2) {
      0: Bell2(q[1], q[2]);
      default: Bell2(q[7], q[8]);
34 };
```

A Sample Code of MQCC which shares many features with OpenQASM



```
1 \\Register and variable declarations
2 greg g[10];
3 creq r[1];
4 fcho c1 = \{0, 1\};
6 \text{ \lange cz} = [0, 1];
5 \text{ fcho c2} = [0, 1];
8 \\Module define
9 module Bell1(q1,q2){
     h(q1);
  cnot(q1, q2);
14 module Bell2(q1, q2){
  case (r[0]) {
   1: x(q1);
   0: pass
   };
h(q1);
     cnot (q1, q2);
23 \Main part of the program
24 choice (c1) {
  0: Bell1(q[1], q[2]);
    1: Bell1(q[7], q[8]);
29 h (q[0]);
30 measure(q[0],r[0]);
31 choice (c2) {
0: Bell2(q[1], q[2]);
default: Bell2(q[7], q[8]);
```

Define CHOICE variables

Free Choice (fcho) c1, c2 $\in \mathbb{Z}$, in certain ranges Limited Choice (lcho) c=1-c1*c2 $\in \mathbb{Z}$

Stitch Many Programs w/ choice variables

choice (c.v) $\{i: P_i\}$

```
\begin{split} n \in \mathbb{N} & i \in \mathbb{Z} \quad r \in \mathbb{R} \quad var \in Vars \\ qreg \in Quantum \ reg. \quad creg \in Classical \ reg. \\ reg ::= qreg \mid creg \\ P \in Program ::= \overrightarrow{D} \ S \\ D \in Declaration ::= RegDecl \mid VarDecl \\ RegDecl ::= \mathbf{qreg} \ qreg; \mid \mathbf{creg} \ creg; \\ VarDecl ::= Free \mid Limit \\ Free ::= \mathbf{fcho} \ var = \{\overrightarrow{i}\}; \mid \mathbf{fcho} \ var = [i_1, i_2]; \\ Limit ::= \mathbf{lcho} \ var = E; \\ E \in VarExp ::= i \mid var \mid E + E \mid E - E \\ \mid E * E \mid E/E \mid (E) \\ S \in Stmt ::= \epsilon \mid O \mid case \mid choice \mid S; S \\ O \in Operation ::= x(\overrightarrow{r}, \overrightarrow{reg}) \\ case ::= \mathbf{case}(creg)\{\overrightarrow{i} : S_i\} \\ choice ::= \mathbf{choice}(var)\{\overrightarrow{i} : S_i\} \end{split}
```

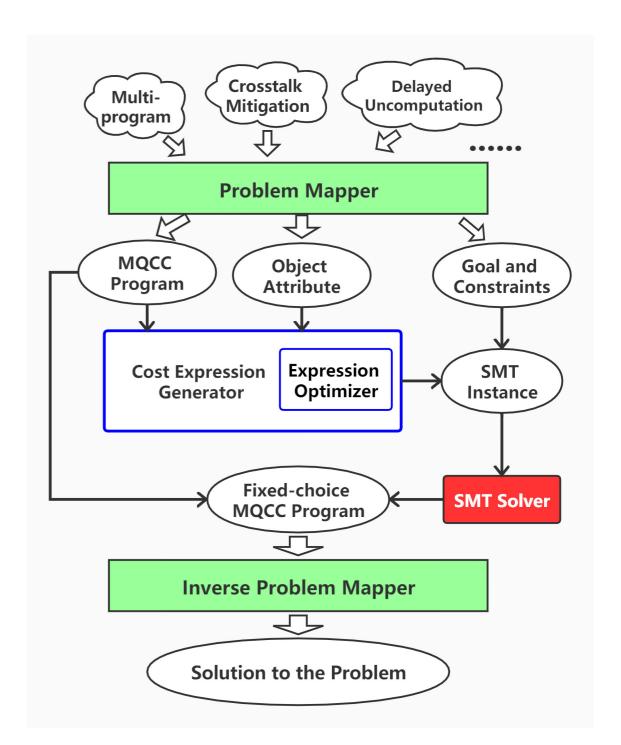
```
1 \\Register and variable declarations
2 greg g[10];
3 creg r[1];
4 fcho c1 = \{0, 1\};
5 fcho c2 = [0, 1];
color cc = [0, 1];
color cc = [0, 1];
color cc = [0, 1];
8 \\Module define
9 module Bell1(q1,q2){
     h(q1);
  cnot(q1, q2);
12 }
14 module Bell2 (q1, q2) {
     case (r[0]) {
    1: x(q1);
   0: pass
   };
    h(q1);
     cnot (q1,q2);
23 \\Main part of the program
24 choice (c1) {
  0: Bell1(q[1], q[2]);
   1: Bell1(q[7], q[8]);
27 : };
29 h (q[0]);
measure(q[0],r[0]);
31 choice (c2) {
0: Bell2(q[1], q[2]);
default: Bell2(q[7], q[8]);
```

A Sample Code of MQCC which shares many features with OpenQASM

Depth: $7\delta_{c_1}^0 \delta_{c_2}^0 + 5\delta_{c_1}^0 \delta_{c_2}^1 + 5\delta_{c_1}^1 \delta_{c_2}^0 + 7\delta_{c_1}^1 \delta_{c_2}^1$

Noise: $0.045\delta_{c_1}^0 + 0.066\delta_{c_1}^1 + 0.027\delta_{c_2}^0 + 0.043\delta_{c_2}^1$

where $\delta_c^i = 1$ iff c = i; otherwise 0



Expressing the Constraints on Costs/Attributes

Express desired goals as objects called Attributes. Thus, any MQCC program is a transformer on attributes.

Precisely, any attribute A is defined by a tuple (*T, empty, op, case, value*) s.t.:

- \bullet T is a data type of the states. A state of type T consists of information needed in the computation of the cost.
- \bullet empty: T is the initial state at the beginning of the program.
- op: $T \times \text{string} \times \overrightarrow{\mathbb{R}} \times \overrightarrow{reg} \to T$ receives a state, an operation's name and its arguments, and generates a new state that merges the old state and the information of the operation.
- case: $T \times \overrightarrow{T} \to T$ receives an old state, a list of states corresponding to each case branch which has merged the corresponding sub-programs' information on the old state, and generates a new state merging the old state and the sub-programs' states.
- value : $T \to \mathbb{R}$ computes the cost of this attribute from the information stored in a state.

choice vars
$$[S]: (Vars \to \mathbb{Z}) \times T \to T$$
 transformer on T

$$S = opID(exps, regs)$$

$$\overline{[S]}(\sigma, s) = op(s, opID, exps, regs)$$

$$\overline{[S_1; S_2]}(\sigma, s) = \overline{[S_2]}(\sigma, \overline{[S_1]}(\sigma, s))$$

$$S = \mathbf{case}(creg)\{\overline{i: S_i}\}$$

$$\overline{[S]}(\sigma, s) = case(s', \overline{[S_i]}(\sigma, s')]_i)$$

$$S = \mathbf{choice}(var)\{\overline{i: S_i}\} \qquad k = \sigma[var]$$

$$\overline{[S]}(\sigma, s) = \overline{[S_k]}(\sigma, s)$$

how transformers evolve over programs



Expressing the Constraints on Costs/Attributes

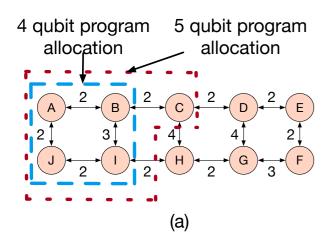
Simple Examples of Attributes

```
Attribute Noise:
                       noise : \mathbb{R}
  T:
  empty():=
                      init s : T, s.noise = 0
                      return s
  value(s : T):= return s.noise
  op (s : T, OpID : str, exps : \overrightarrow{\mathbb{R}}, regs : \overrightarrow{Reg}) :=
                s.noise += calNoise(OpId, exps, regs)
                       return s
  case (s : T, group : Vector of T) :=
                       s.noise = \max \{n.\text{noise} | n \in \text{group}\}
                       return s
Attribute Depth:
  T: dep : Map of Reg \to \mathbb{N}
  empty():= init s : T, s.dep = \emptyset
               return s
  value(s : T):= return (max s.dep.values)
  op (s : T, OpID : str, exps : \overline{\mathbb{R}}, regs : \overline{Reg}) :=
                 share = s.dep.keys \cap regs
                 next = \max \{s.dep[i] | i \in share\} + 1
                 for i \in regs: s.dep.update(i, next)
                 return s
  case (s : T, group : Vector of T) :=
                 all = \bigcup_{n \in \text{group}} \text{ n.dep.keys}
    s.dep = \{(k, \max \{n.dep[k] \mid n \in group\}) \mid k \in all\}
                 return s
```

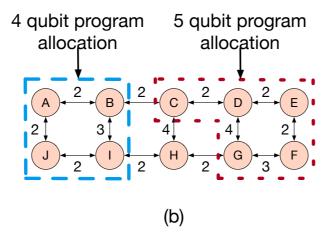
All experiments performed on IBMQ machines

Case Study

Multi-Programming (MICRO 2019):



Competing Goals:Depth vs High-quality Qubits



Probability of Successful Trial

isolated

Sequential: always high-quality qubits

MQCC MQCC

Multi-Programming with MQCC

Recover some essential ideas of MICRO 2019 while ignoring others.

```
qreg q[10];
creg r[1];

module Bell1(q1,q2){
    h(q1);
    cnot(q1, q2);
}

module Bell2(q1, q2){
    case (r[0]){
        1: barrier(q1);
        x(q1);
        0: pass
    }
    h(q1);
    cnot(q1,q2);
}
```

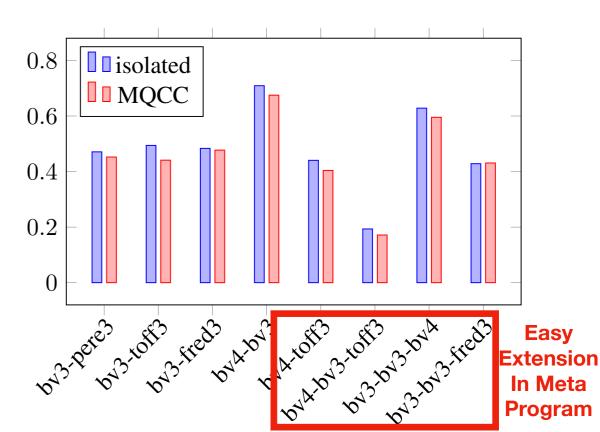
```
fcho c1 = {0, 1};
fcho c2 = [0, 1];

choice (c1){
    0: Bell1(q[1], q[2]);

    1: Bell1(q[7], q[8]);
}

choice (c2){
    0: Bell2(q[1], q[2]);

    1: Bell2(q[7], q[8]);
}
```



Multi-Tasks over Simple Quantum Algorithms

Group A-B contains two applications A and B. Similarly A-B-C.

A success trial if all applications in the group are successful.

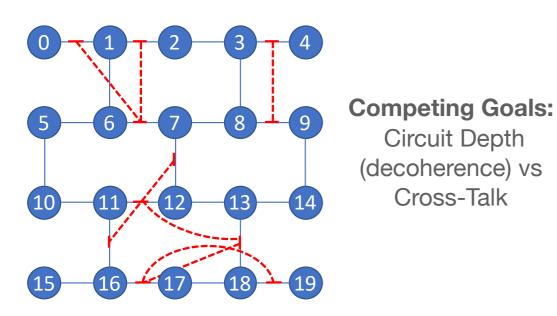
Execute 8192 trials on IBMQ Rochester for each group.

Comparable results to MICRO 2019 for this part.

Application	Description	Qubits	# of gates	# of CNOTs
bv3	Bernstein-Vazirani [3]	3	8	2
bv4	Bernstein-Vazirani [3]	4	11	3
Toff3	Toffoli gate	3	15	6
Fred3	Fredkin gate	3	17	8
Pere3	Peres gate	3	16	7

Case Study

Cross-talk: (Xtalk - ASPLOS 2020)



```
module cnotb(c,q1,q2) {
      choice (c) {
          0: cnot(q1,q2);
          1: barrier(q1,q2);
          cnot(q1,q2);
      }
}
```

use "barrier" to control the order of the gates

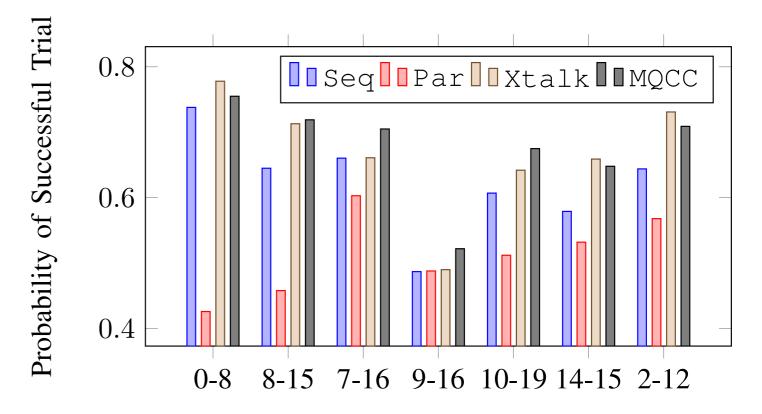
Benchmark Test:

CNOT 15 8 = SWAP 15 16; SWAP 16 11; SWAP 8 7I SWAP 7 12; CNOT 11 12

Execute 8192 trials on IBMQ Boeblingen for SWAP circuits connecting a-b

```
Seq running all instructions serially
```

Par maximize the parallel execution, default in Qiskit



Benchmark over SWAP circuits connecting a-b on IBM Boeblingen

```
Attribute Crosstalk
        dep : Map of Reg 	o \mathbb{N}
        rep : Map of \mathbb{N} \to \mathbf{Set} of (\operatorname{str} \times Req')
                     init s:T, s.dep = \emptyset, s.rep = \emptyset
  empty ():=
                     return s
  value (s : T):= return calCross(rep)
  op (s : T, opID : str, exps : \overrightarrow{\mathbb{R}}, regs : \overrightarrow{Reg}) :=
        if opId == "barrier" :
          cur = max s.dep.values
          for i∈regs: s.dep.update(i, cur)
        else:
          share = s.dep.keys \cap regs
          next = max \{s.dep[i] \mid i \in share\} + 1
          s.rep[next].insert( (OpID, regs) )
          for i∈regs: s.dep.update(i, next)
        return s
  case (s : T, group : Vector of T) :=
        all = [] n.dep.keys, n∈group
        s.dep = \{(k, \max \{n.dep[k]|n\in group\})| k\in all\}
  s.rep = \{(k, \bigcup_{n \in \text{group}} n.\text{rep}[k]) | \exists u \in \text{group}, k \in u\}
```

Case Study: Multi-Programming + Cross-Talk

Optimizing Goal:

EASY implementation in MQCC

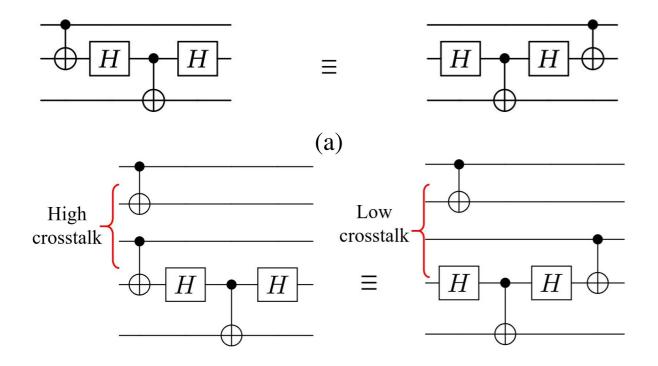
Noise + Decoherence + Crosstalk

Seq Sequential: always high-quality qubits. but larger depth (decoherence)

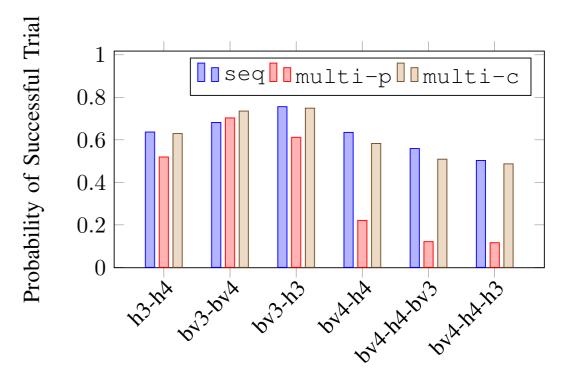
multi-p Multi-programs without considering crosstalk short depth, but large crosstalk errors

multi-c Multi-programs with crosstalk short depth and large successful probability

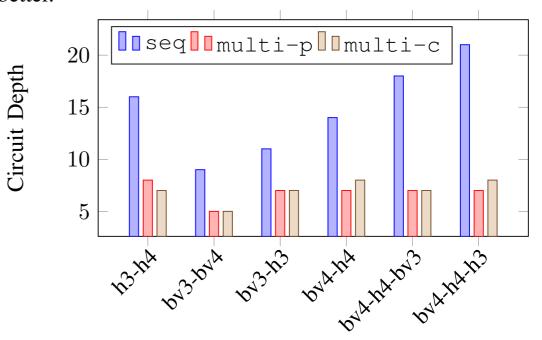
More Optimization w/ MQCC



All experiments performed on IBMQ machines



(a) Probability of Successful Trial. Here higher PST is better.



(b) Circuit Depth. Here lower circuit depth is better.

Case Study: Cost-Effective Uncomputation

Recovering one idea from SQUARE (ISCA 2020)

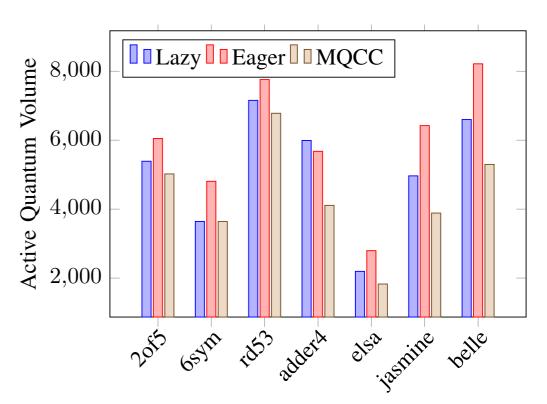
Strategic Quantum Ancilla Reuse for Modular Quantum Programs

Deciding the point to uncompute for ancilla reuse

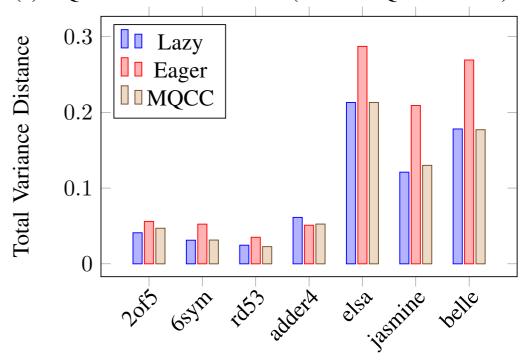
```
fcho c1,c2 = {0,1};
lcho ct = 1 - c1*c2;

foo1(c1,...);
foo2(c2,...);
choice (ct) {
    0: pass \\No uncomputation
    1: uncompute code \\Do uncomputation
}
```

Name	Discription	Gate Number	Qubits
2of5	Output is 1 if number of 1s in its input equals two.	1528	8
6sym	Function with 6 inputs and 1 output.	1620	11
rd53	Input weight function with 5 inputs and 3 outputs.	1849	10
adder4	4-bit in-place controlled-addition.	1748	12
elsa	Heavy workload and shallowly nested synthetic function.	256	14
jasmine	Shallowly nested synthetic function.	604	11
belle	Light workload and deeply nested synthetic function.	768	9



(a) AQV of the benchmarks. (Lower AQV is better.)



(b) Realistic noise simulation using IBM Qiskit Aer simulator. (Lower total variation distance is better.)

Thank You!





MQCC:

- github/sqrta/MQCC