

Quantum Cryptanalysis in the RAM Model: Claw-Finding Attacks on SIKE

Samuel Jaques and John M. Schanck

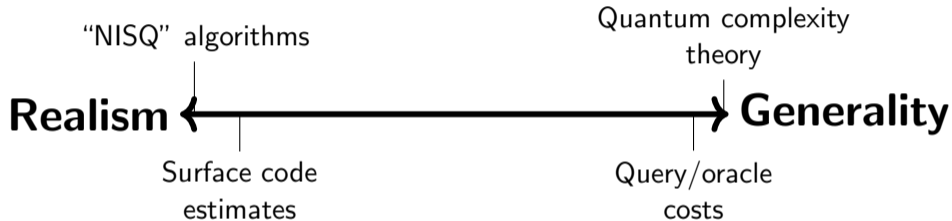


Models of quantum computers

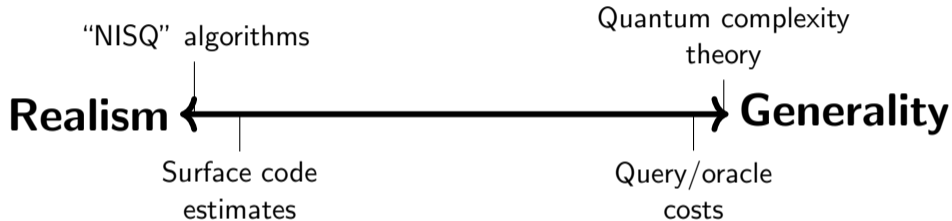
- NIST is working on post-quantum public key standards
- This requires **quantum cryptanalysis**
- This requires models of quantum computers

How do you imagine a quantum computer?

Quantum cost analysis

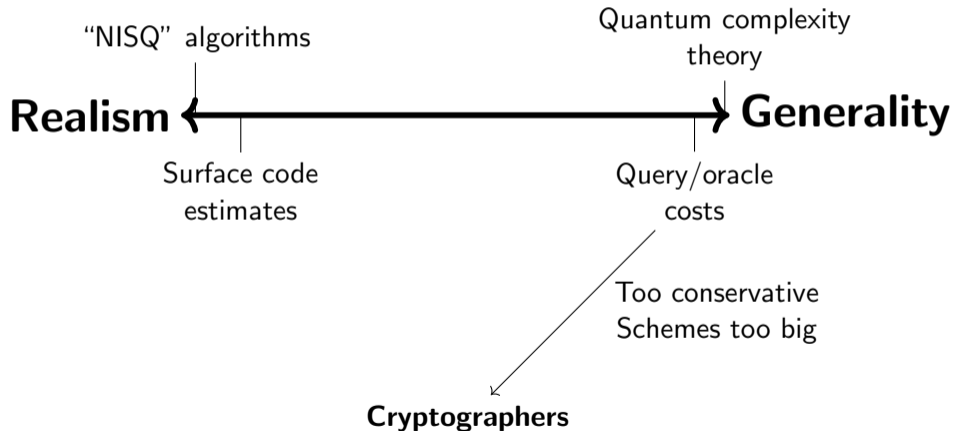


Quantum cost analysis

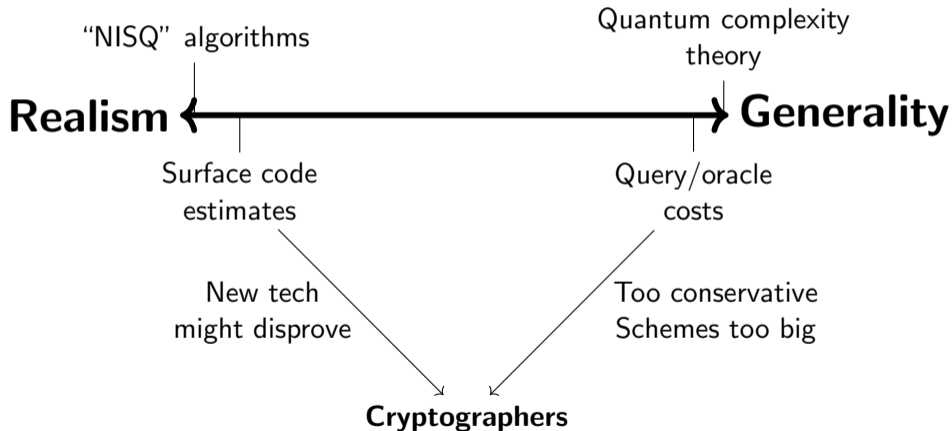


Cryptographers

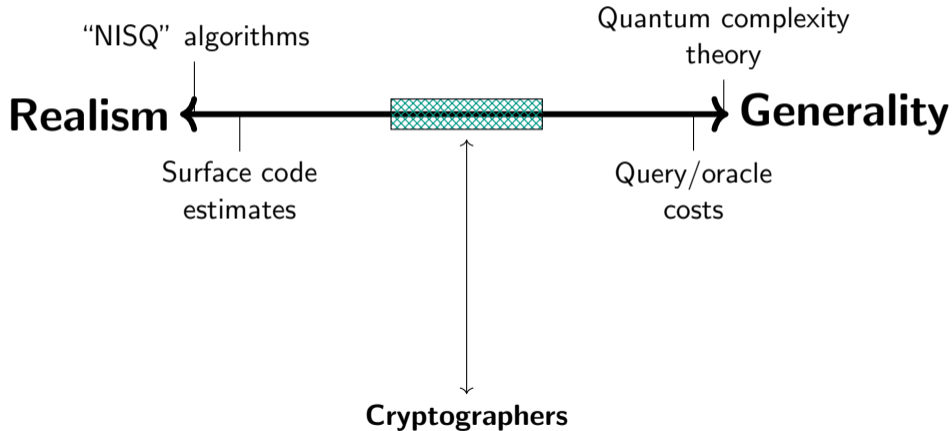
Quantum cost analysis



Quantum cost analysis



Quantum cost analysis



Outline

- 1 Motivation
- 2 Memory peripheral framework
- 3 Cost models
- 4 Analysis of SIKE
- 5 Summary

Goal 1: Fairly compare classical and quantum resources

How do we compare a quantum bit of security to a classical bit of security?

How do we cost mixed classical/quantum algorithms like Kuperberg's sieve?

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How do we cost mixed classical/quantum algorithms like Kuperberg's sieve?

Previous work: Analysis of Brassard-Høyer-Tapp (BHT)

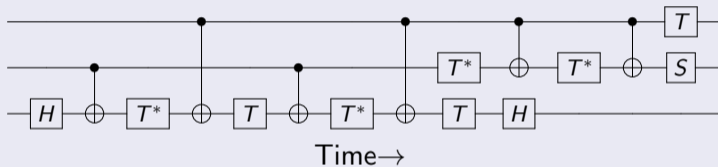
- BHT provided a quantum collision-finding algorithm with quantum access to classical memory.
- Bernstein argued van Oorschot-Wiener is more efficient after fully accounting for memory costs.

Brassard, Høyer, Tapp. 1997. Quantum Algorithm for the Collision Problem

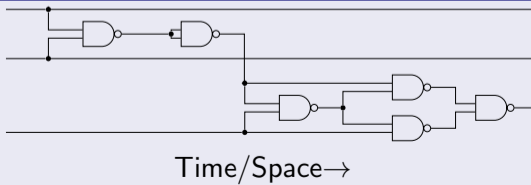
Bernstein. 2009. Cost analysis of hash collisions: Will quantum computers make SHARCS obsolete?

Goal 2: View gates as processes

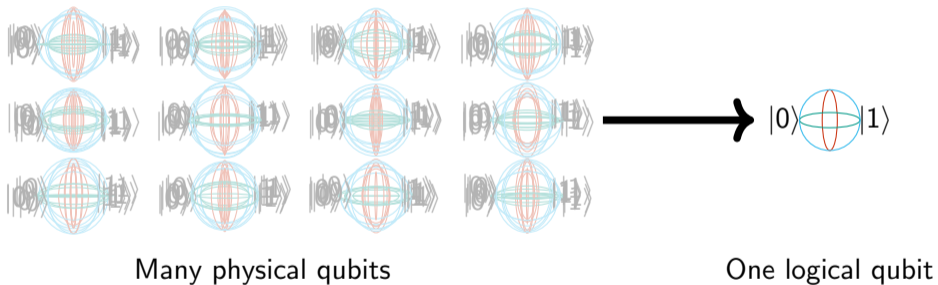
Quantum



Classical



Goal 3: Include error correction



Memory peripheral framework

Main Idea

Model computation as “memory” acted on by a “memory controller”.

Examples:

- Turing machine: head + tape
- RAM: CPU + random access memory
- Quantum circuit: Random access machine + qubits

Premises:

- 1 **Memory** is a physical system that changes over time
- 2 A **memory controller** interacts with a memory
- 3 The **cost** of a computation is the number of interactions

Premise 1: Memory is a physical system

Free evolution

Caused by:

- Noise
- Ballistic computation

Costly evolution

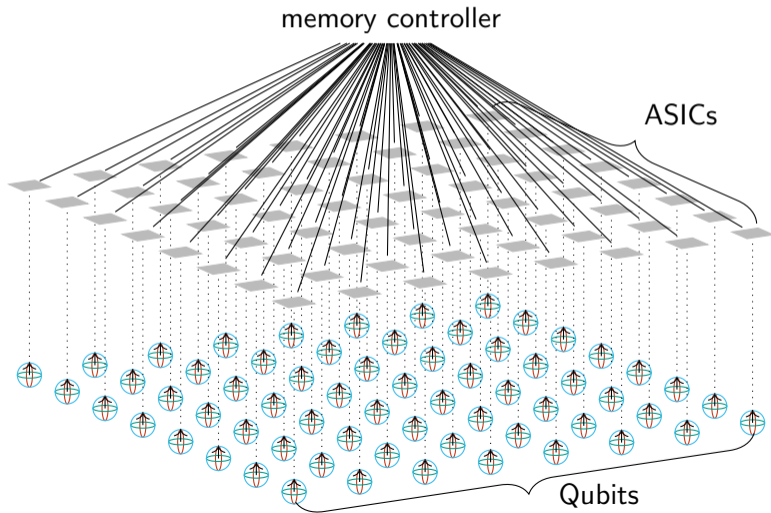
Caused by the controller.

We model a quantum computer as a **parallel random access machine** with new instructions for quantum gates

- e.g.: apply gate x to qubit y at time t

Result: quantum algorithms are classical programs

Premise 2: Memory controller



Premise 3: Cost

The **cost** of a computation is the number of interactions.

- We ignore the construction cost
- We focus on the cost to the controller

There are opportunity costs: What else could the controller do?

Cost models

We provide physical justifications for two cost models: **G-cost** and **DW-cost**.

Both are qubit memories with a standard universal gate set (Clifford + T).

Differences:

- *G*-cost: **Passive** error correction.
- *DW*-cost: **Active** error correction.

Error correction

Passive/Non-volatile memory

To preserve: keep cool.

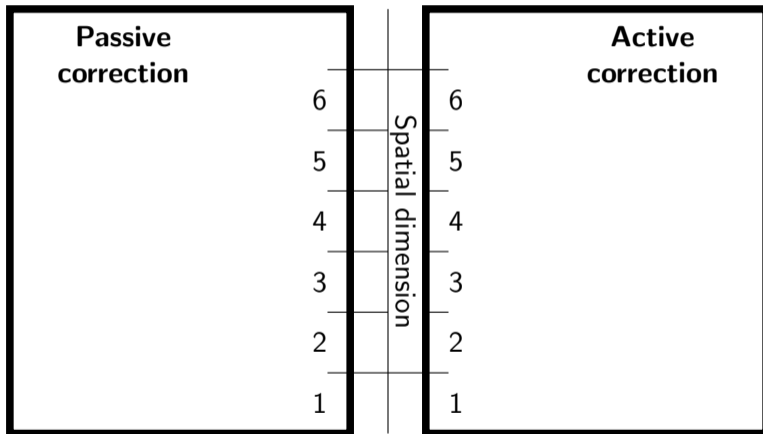
- Paper
- Magnetic discs

Active/Volatile memory

To preserve: continuously refresh.

- DRAM
- Surface codes (quantum)

Quantum error correction theory

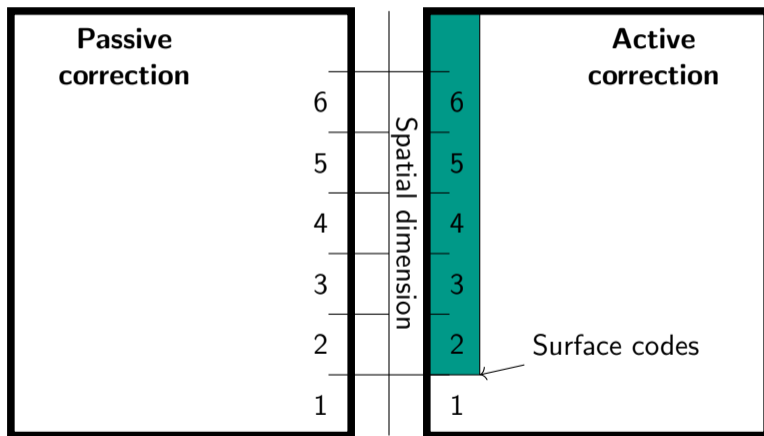


[1] Bravyi and Terhal. 2009. A no-go theorem for a two-dimensional self-correcting quantum memory based on stabilizer codes.

[2] Kitaev. 2003. Fault-tolerant quantum computation by anyons.

Dennis, Kitaev, Landahl, Preskill. 2002. Topological Quantum Memory.

Quantum error correction theory

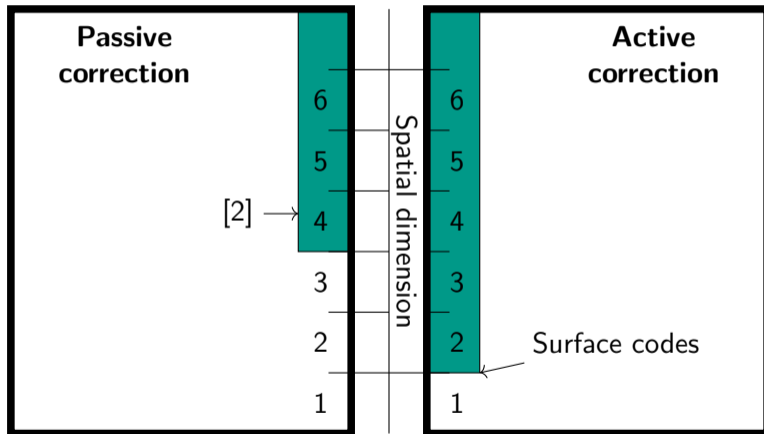


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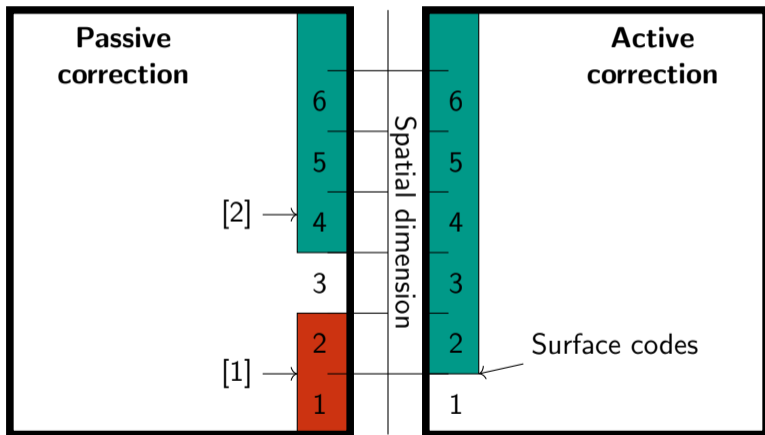


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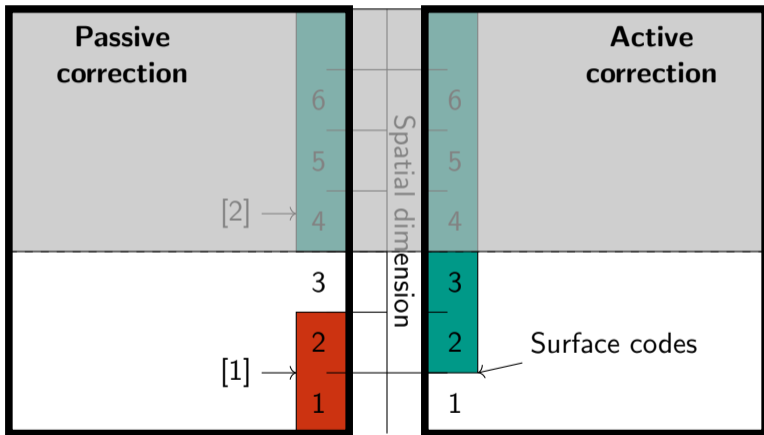
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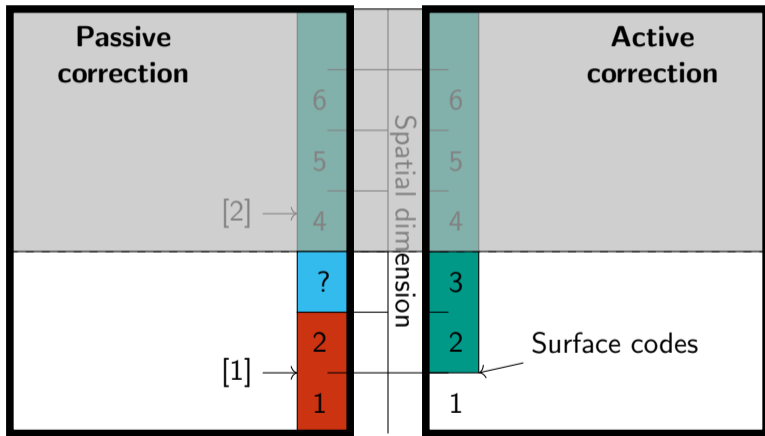
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Costs

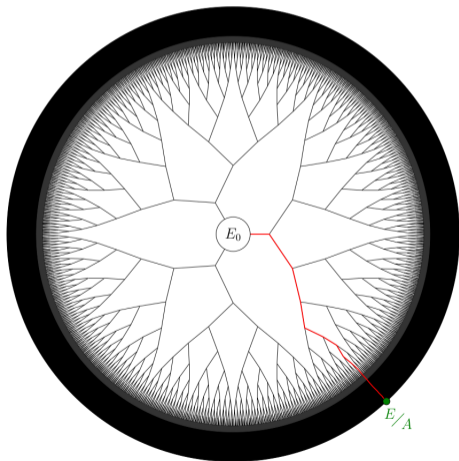
G-cost

- Assumption: **Passive** error correction.
(Physical, not just technological, assumption)
- Cost: 1 RAM operation per gate
- Total cost: Number of gates ("*G*")

DW-cost

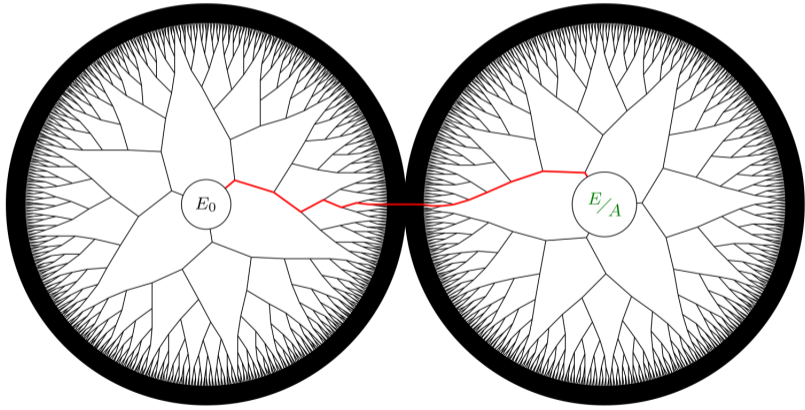
- Assumption: **Active** error correction.
- Cost: 1 RAM operation per qubit per time step
- Total cost: Depth \times Width ("*DW*")

Analysis of SIKE



- E_0 is public parameter, E/A is public key
- Parameterized by a large prime p (e.g. $p \approx 2^{434}$)
- Red path is secret key (length $\log p/2$)

Meet-in-the-middle



Tani's collision-finding algorithm

To find a collision between two functions $f : X \rightarrow S$ and $g : Y \rightarrow S$:

- Random walk on two Johnson graphs: one over X , the other over Y
- Check for collisions at each step
- Make it quantum!

Johnson graph over X

Vertices: R -element subsets of a fixed set X .

Vertices u and v are adjacent iff $|u \cap v| = R - 1$.

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Query-optimal parameters:

$$R = \# \text{ queries} = \text{time}$$

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Johnson graph over X

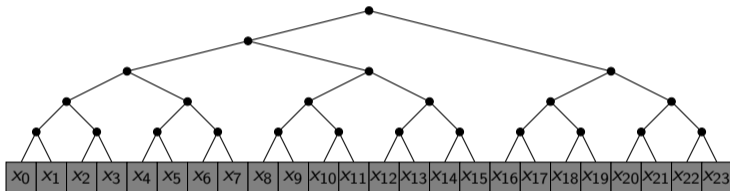
Vertices: R -element subsets of a fixed set X .

Vertices u and v are adjacent iff $|u \cap v| = R - 1$.

Query-optimal parameters to attack SIKE:

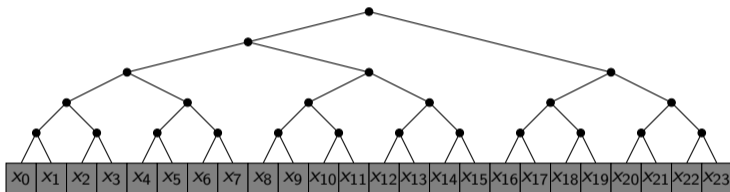
$$R = \# \text{ queries} = \text{time} = p^{1/6+o(1)}$$

Memory access



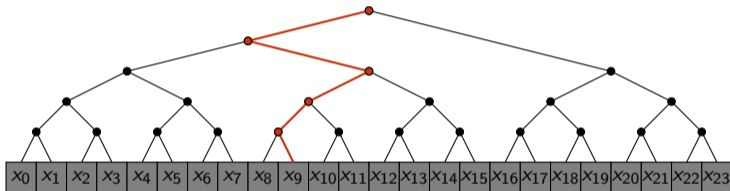
Memory access

Classical Query: 9



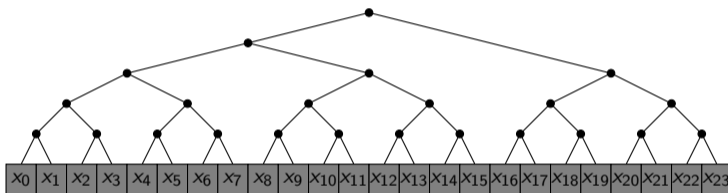
Memory access

Classical Query: 9



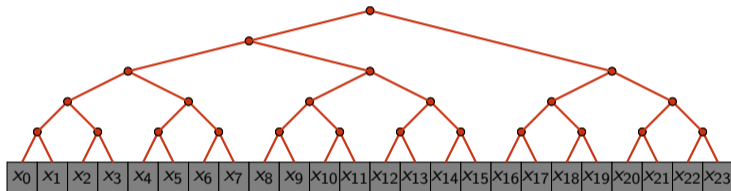
Memory access

Quantum Query: 024681012



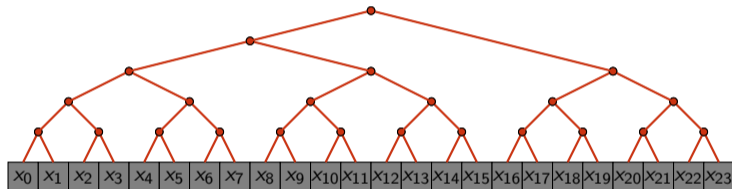
Memory access

Quantum Query: `0x40000000`



Memory access

Quantum Query: `0249112123`



Analogy for Cryptographers

- Any physical “side channel” leaks information
- Any leaked information decoheres (destroys) the state
- Controller must implement circuits for all possible inputs

Memory costs

For N bits of random-access quantum memory:

Idle memory

- G -cost: Free
- DW -cost: $O(N)$ RAM ops per time step

Random access

- G -cost: $O(N)$ RAM ops
- DW -cost: $O(N \log N)$ RAM ops

Johnson vertex data structure

History independence

For quantum interference between random walk paths, the representation of a vertex must be independent of the path taken.

Johnson vertex data structure

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History-dependent:

- Binary tree as linked list

Johnson vertex data structure

History independence

For quantum interference between random walk paths, the representation of a vertex must be independent of the path taken.

History-dependent:

- Binary tree as linked list

History-independent:

- Quantum radix tree: superposition over all layouts
- Sorted array: physically in order

Johnson vertex insertion

Idea: We already pay $O(N)$ for memory access, so pay $O(N)$ to physically sort array:

A' :

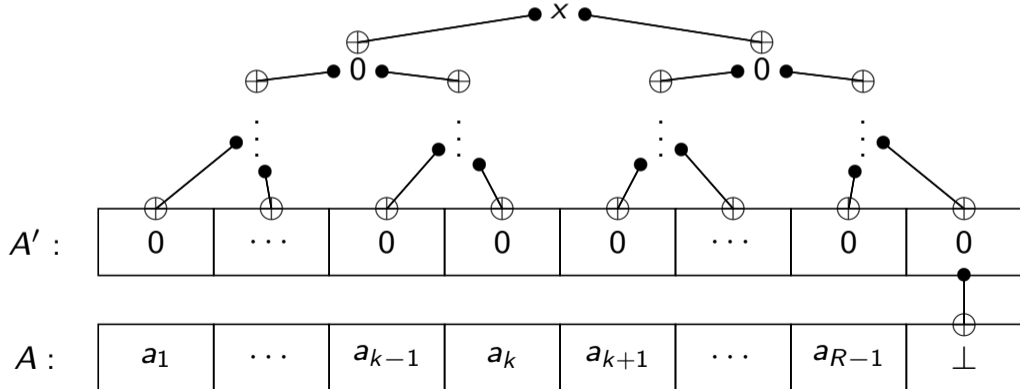
0	...	0	0	0	...	0	0
---	-----	---	---	---	-----	---	---

A :

a_1	...	a_{k-1}	a_k	a_{k+1}	...	a_{R-1}	\perp
-------	-----	-----------	-------	-----------	-----	-----------	---------

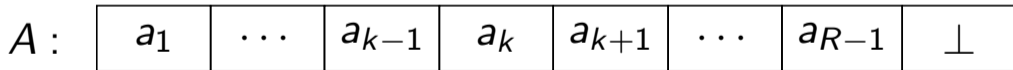
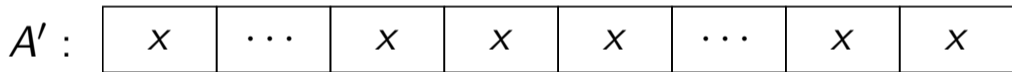
Johnson vertex insertion

1. "Fan out" an input x



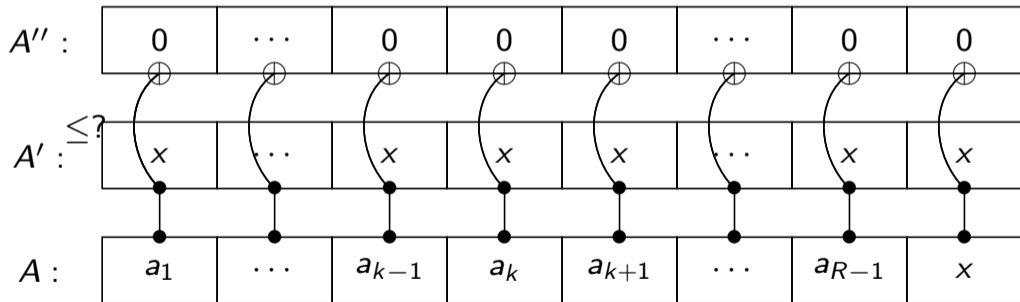
Johnson vertex insertion

1. “Fan out” an input x



Johnson vertex insertion

2. Compare all elements simultaneously



Johnson vertex insertion

2. Compare all elements simultaneously

$$A'' :$$

0	...	0	1	1	...	1	1
---	-----	---	---	---	-----	---	---

$$A' :$$

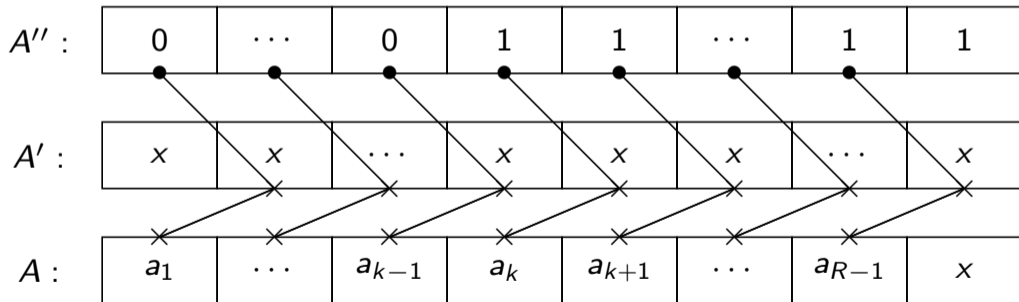
x	...	x	x	x	...	x	x
---	-----	---	---	---	-----	---	---

$$A :$$

a_1	...	a_{k-1}	a_k	a_{k+1}	...	a_{R-1}	x
-------	-----	-----------	-------	-----------	-----	-----------	---

Johnson vertex insertion

3. Conditionally swap “up”



Johnson vertex insertion

3. Conditionally swap “up”

$$A'' :$$

0	...	0	1	1	...	1	1
---	-----	---	---	---	-----	---	---

$$A' :$$

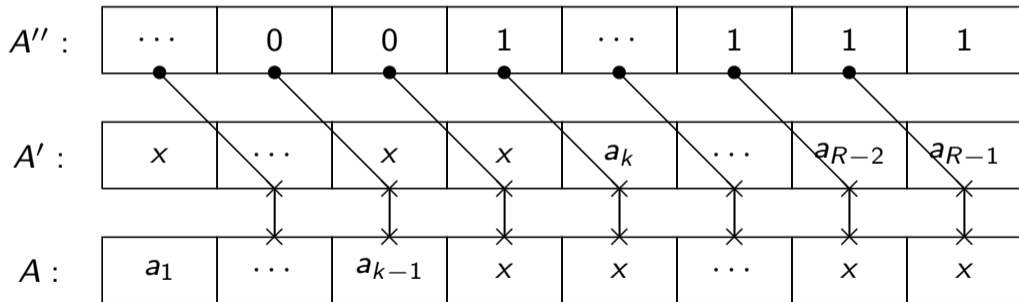
x	...	x	x	a_k	...	a_{R-2}	a_{R-1}
-----	-----	-----	-----	-------	-----	-----------	-----------

$$A :$$

a_1	...	a_{k-1}	x	x	...	x	x
-------	-----	-----------	-----	-----	-----	-----	-----

Johnson vertex insertion

4. Conditionally swap “down”



Johnson vertex insertion

4. Conditionally swap “down”

$$A'' :$$

...	0	0	1	...	1	1	1
-----	---	---	---	-----	---	---	---

$$A' :$$

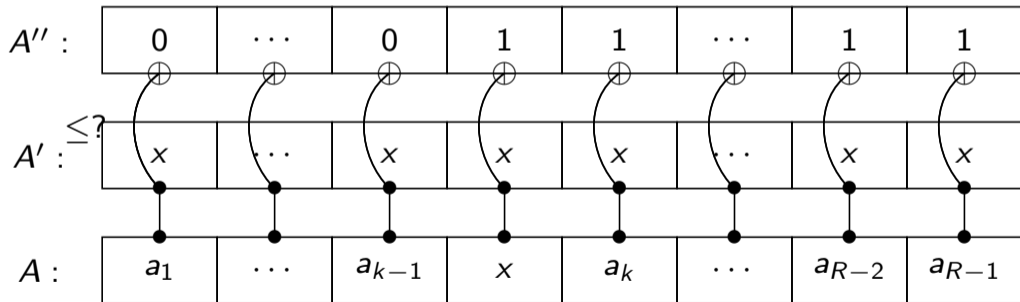
x	...	x	x	x	...	x	x
---	-----	---	---	---	-----	---	---

$$A :$$

a_1	...	a_{k-1}	x	a_k	...	a_{R-2}	a_{R-1}
-------	-----	-----------	---	-------	-----	-----------	-----------

Johnson vertex insertion

5. Clear comparison bit



Johnson vertex insertion

5. Clear comparison bit

A'' :

0	...	0	0	0	...	0	0
---	-----	---	---	---	-----	---	---

A' :

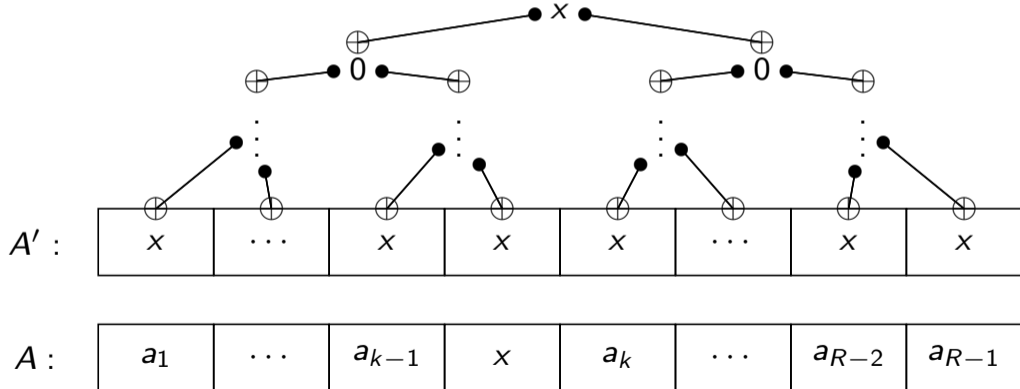
x	...	x	x	x	...	x	x
---	-----	---	---	---	-----	---	---

A :

a_1	...	a_{k-1}	x	a_k	...	a_{R-2}	a_{R-1}
-------	-----	-----------	---	-------	-----	-----------	-----------

Johnson vertex insertion

7. Clear fan-out



Johnson vertex insertion

8. Insertion complete

A' :

0	...	0	0	0	...	0	0
---	-----	---	---	---	-----	---	---

A :

a_1	...	a_{k-1}	x	a_k	...	a_{R-2}	a_{R-1}
-------	-----	-----------	-----	-------	-----	-----------	-----------

Costs of Tani's algorithm for SIKE

Previous analyses focused on the $p^{1/6}$ query cost of Tani's algorithm.

Using the Johnson vertex data structure, we find the SIKE secret at cost:

	Gates	Depth	Width	DW
Tani (query-optimal)	$p^{1/3+o(1)}$	$p^{1/6+o(1)}$	$p^{1/6+o(1)}$	$p^{1/3+o(1)}$

$$2^{434} \leq p \leq 2^{951}$$

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Tani (DW -optimal)	$p^{1/4+o(1)}$	$p^{1/4+o(1)}$	$p^{o(1)}$	$p^{1/4+o(1)}$

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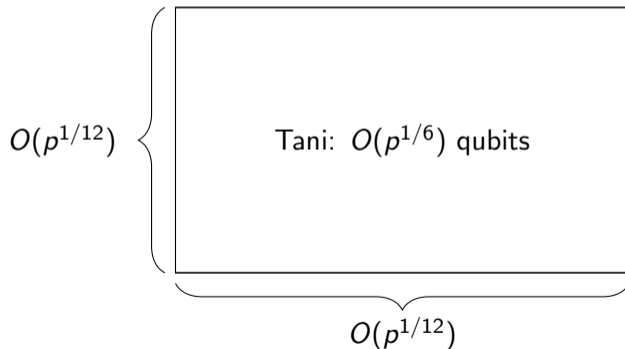
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Grover (G -optimal)	$p^{1/4+o(1)}$	$p^{1/4+o(1)}$	$p^{o(1)}$	$p^{1/4+o(1)}$

$$2^{434} \leq p \leq 2^{951}$$

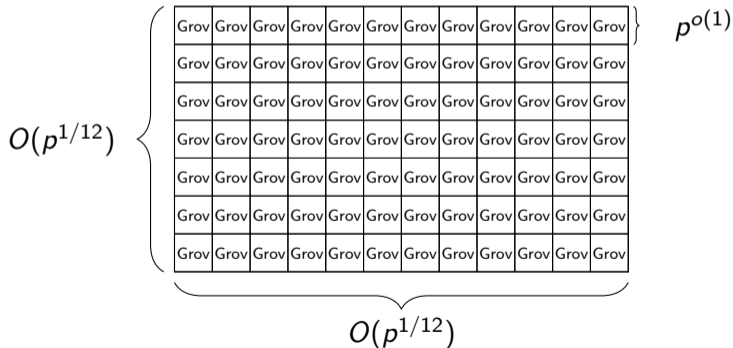
Comparison with parallel Grover

The classical controller can apply gates to every qubit to run Tani's algorithm. It could instead group them together and run Grover's search algorithm.



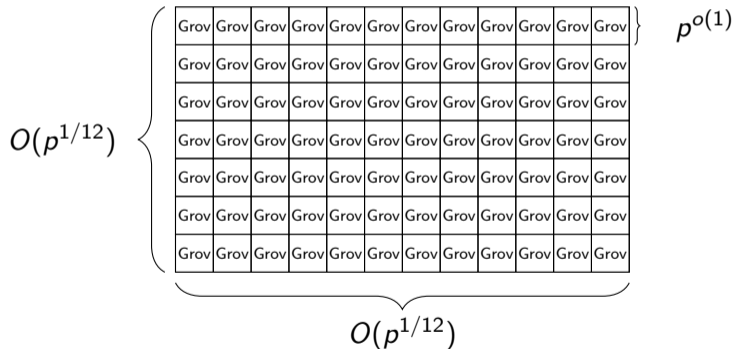
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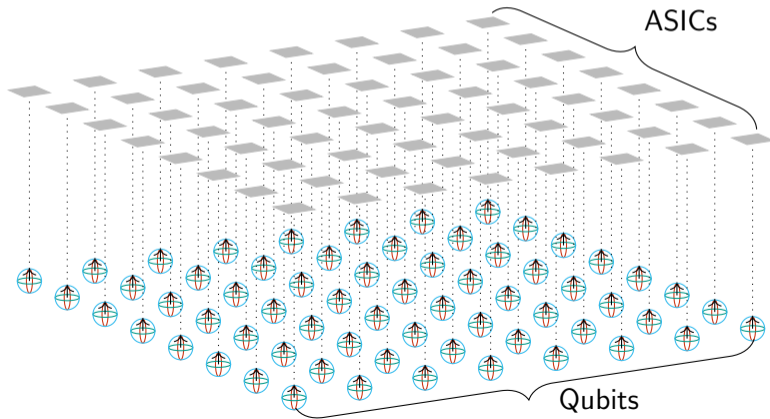
Comparison with parallel Grover

$O(p^{1/6})$ copies of Grover finds isogeny in time $O(p^{1/6})$.



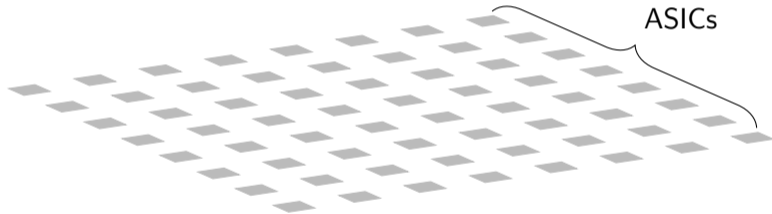
Comparison with van Oorschot-Wiener

- Time/query-optimal Tani has $O(p^{1/6})$ classical control processors.
- We could reprogram these to run van Oorschot-Wiener (VW)



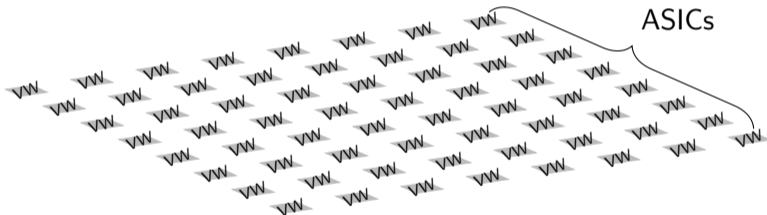
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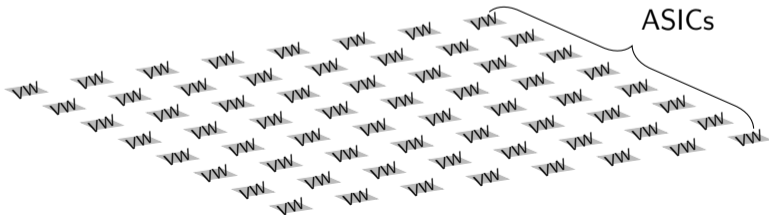


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Conclusion

$O(p^{1/6})$ parallel instances of van Oorschot-Wiener find isogeny in time $O(p^{1/8})$, faster than the quantum algorithms.



Memory peripheral framework

- 1 **Memory** is a physical system that changes over time
- 2 A **memory controller** interacts with a memory
- 3 The **cost** of a computation is the number of interactions

Conclusions

- In a quantum computer, qubits are a peripheral of a classical computer.
- Quantum memory access has a linear gate cost.
- Active error correction gives cost to the identity gate.
- SIKE is more secure than previously thought.