

# Exploiting preprocessing to break $\mathcal{MQ}$ system parameters

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## Quantum computers and cryptography

The well-known dangers of quantum computers fit into the two standard camps of algorithmic cryptanalysis.

- Quantum computers provide an exponential speedup for currently used public-key cryptosystems via Shor's algorithms for the factoring and discrete-log problems.
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- Impact: start again with new hardness assumptions.
- Quantum computers provide a square-root speedup in query-complexity for brute force search methods.
- Impact: when search is thought to be the best approach to solving a problem, such as for breaking AES, we double the size of the key. 128-bit security becomes 64-bit security, and so forth...

# Quantum cryptanalysis

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↑  
This paper



# Overview

The MQ-hardness assumption

Quantum computing 101

Quantum search and cost

Quantum circuits and the  $\mathcal{MQ}$  evaluation oracle

Our adaptation

## **The MQ-hardness assumption**

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## One of many candidates

Many post-quantum hardness assumptions we could choose instead of factoring or the discrete-log problem.

- Lattice problems
- Hashing problems
- Isogenies on supersingular curves
- Code-based cryptography
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# The hard problem

## Definition (The Multivariate Quadratic ( $\mathcal{MQ}$ ) problem)

Let  $p^{(1)}, \dots, p^{(m)} \in \mathbb{F}_q[x_1, \dots, x_n]$ , where  $\mathbb{F}_q$  is a finite field of size  $q$  and each  $p^{(i)}$  is of degree two.

The Multivariate Quadratic ( $\mathcal{MQ}$ ) problem is to find an  $\bar{x} = (x_1, \dots, x_n)$  with  $x_i \in \mathbb{F}_q$  such that  $p^{(i)}(\bar{x}) = 0$  for  $i = 1, \dots, m$ .

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but such an approach ignores structure in the problem.

We'll be looking only at the  $q = 2$  (binary  $\mathcal{MQ}$ ) case from here on.

## Why is it important?

For our purposes: Hidden Field Equations (HFE) cryptosystems.

Interesting topic — but too little time in this talk.

Target for cryptanalysis: The Gui  $\mathcal{MQ}$  signature scheme.

Public-key of a  $\text{Gui}(n, D, a, v, k)$   $\mathcal{MQ}$  signature scheme is the  $\mathcal{MQ}$  map

$$\mathcal{P} : \mathbb{F}_2^{n+v} \longrightarrow \mathbb{F}_2^{n-a} \quad (2)$$



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Too little time to detail the workings of Gui, but:

- Signing a message  $\equiv$  inverting  $\mathcal{P}$  a total of  $k$  times.
- Easy to invert if we know structure of  $\mathcal{P} = L_2 \circ \Phi^{-1} \circ \mathcal{F} \circ \Phi \circ L_1$ .
- $\mathcal{F} \in \mathbb{E}[X]$  where  $[\mathbb{E} : \mathbb{F}_2] = n$  and  $\text{degree}(\mathcal{F}) = D$ .
- Easier to solve using Gröbner/XL techniques than random  $\mathcal{MQ}$ .
- Search based methods do not exploit this structure.

# Classical / quantum methods for solving the $\mathcal{MQ}$ problem.

## Leading classical methods

- Exhaustive search methods.
- BooleanSolve.
- Gröbner bases techniques:  $XL/F_4/F_5$ .

## Leading quantum methods

- Quantum search for small parameter sizes.
- QuantumBooleanSolve (BooleanSolve/Grover hybrid).
- GroverXL (XL/Grover hybrid).

# Quantum computing 101

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# Multiple layers of abstraction for quantum computing

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- Abstract unitary operators on a Hilbert space
- Logical quantum circuit level
- Fault-tolerant level
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Unitary operators — how we describe algorithms.

Quantum circuits — how we implement unitary operators.

# Quantum states and the computational basis

An  $n$ -qubit quantum state can be written in the *computational basis* as

$$|\psi\rangle = \sum_{x \in \{0,1\}^n} \alpha_x |x\rangle \quad \text{where } \alpha_x \in \mathbb{C} \quad \text{and} \quad \sum_{x \in \{0,1\}^n} |\alpha_x|^2 = 1.$$

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Quantum algorithm design: evolving an initial quantum state to one whose amplitudes which encode useful information are amplified.

## Quantum states and quantum gates I

Quantum states can be viewed as vectors of coefficients.

$$|\psi\rangle = \alpha_{00} |00\rangle + \alpha_{01} |01\rangle + \alpha_{10} |10\rangle + \alpha_{11} |11\rangle = \begin{bmatrix} \alpha_{00} \\ \alpha_{01} \\ \alpha_{10} \\ \alpha_{11} \end{bmatrix} \quad (3)$$

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Processes which evolve one quantum state to another in a period of continuous time are *unitary operators* acting upon these vectors.

$$U |\psi_{t_0}\rangle = |\psi_{t_1}\rangle \quad \text{where } U^\dagger U = U U^\dagger = I \quad (4)$$

Note:  $U^\dagger = (U^T)^*$  — the conjugate transpose.

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This connection to linear algebra both simplifies and complicates the implementation of Grover's algorithm.

## Quantum states and quantum gates III

A key component of Grover is the concept of the *quantum oracle*, a unitary operator defined by a boolean function  $F : \{0, 1\}^n \rightarrow \{0, 1\}$ .

$$F(x) = \begin{cases} 1 & \text{if } x \text{ is a target} \\ 0 & \text{otherwise} \end{cases} \quad S_F |x\rangle = \begin{cases} -|x\rangle & \text{if } F(x) = 1 \\ |x\rangle & \text{if } F(x) = 0 \end{cases} \quad (5)$$

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*Linearity* of unitary operators reduces the design of quantum oracles to the problem of implementing a circuit that is correct on bitstrings

$$S_F \sum_{x \in \{0,1\}^n} \alpha_x |x\rangle = \sum_{x \in \{0,1\}^n} \alpha_x S_F |x\rangle \quad (6)$$

## Quantum states and quantum gates IV

A unitary operator implementing  $F : \{0, 1\}^n \longrightarrow \{0, 1\}$  on bitstrings is enough to realise  $S_F$ .

Say we have the unitary  $U_F$

$$U_F |x\rangle |y\rangle \mapsto |x\rangle |y \oplus F(x)\rangle \quad (7)$$



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then replacing  $y$  with the state (where  $H$  is the Hadamard gate)

$$H |1\rangle = \frac{|0\rangle - |1\rangle}{\sqrt{2}} \quad (8)$$

gives us

$$U_F |x\rangle \left( \frac{|0\rangle - |1\rangle}{\sqrt{2}} \right) \mapsto |x\rangle \left( \frac{|0 \oplus F(x)\rangle - |1 \oplus F(x)\rangle}{\sqrt{2}} \right) = (-1)^{F(x)} |x\rangle \left( \frac{|0\rangle - |1\rangle}{\sqrt{2}} \right).$$

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Recall the universal gate set for boolean circuits:  $\{\wedge, \neg, \oplus\}$ .

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# Reversibility

Implementations of boolean circuits are required to be *reversible* because of the unitary condition

$$U^\dagger U = UU^\dagger = I. \tag{12}$$

Impact: all boolean circuits must implement permutations.

Use of ancillae qubits is crucial for efficient realisation.

## **Quantum search and cost**

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## Grover's quantum search algorithm and query complexity I

Consider the boolean function  $F$  with a domain of size  $N = 2^n$

$$F : \{0, 1\}^n \longrightarrow \{0, 1\} \tag{13}$$

with the promise that  $M = |F^{-1}(1)|$ .

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- Given a classical circuit implementing  $F$ :  $\mathcal{O}\left(\frac{N}{M}\right)$  queries by exhaustive search.
- Given a quantum circuit implementing  $F$  :  $\mathcal{O}\left(\sqrt{\frac{N}{M}}\right)$  queries by Grover's algorithm.

## Grover's quantum search algorithm and query complexity II

Let  $F : \{0, 1\}^n \longrightarrow \{0, 1\}$ ,  $N = 2^n$  and  $M = |F^{-1}(1)|$ .

Given the unitary operators acting on  $n$ -qubits

$$S_0 |x\rangle = \begin{cases} -|x\rangle & \text{if } |x\rangle = |0\rangle \\ |x\rangle & \text{if } |x\rangle \neq |0\rangle \end{cases} \quad S_F |x\rangle = \begin{cases} -|x\rangle & \text{if } F(x) = 1 \\ |x\rangle & \text{if } F(x) = 0 \end{cases}, \quad (14)$$

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$$G = -H^{\otimes n} S_0 H^{\otimes n} S_F \quad (15)$$

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4. Perform a measurement in the computational basis.

## Grover's quantum search algorithm and query complexity III

If we define the normalised superpositions

$$|Good\rangle = \frac{1}{\sqrt{M}} \sum_{\substack{x \in \{0,1\}^n \\ x:F(x)=1}} |x\rangle \quad \text{and} \quad |Bad\rangle = \frac{1}{\sqrt{N-M}} \sum_{\substack{x \in \{0,1\}^n \\ x:F(x)=0}} |x\rangle \quad (16)$$

and the angle  $\theta = \arcsin\left(\sqrt{\frac{M}{N}}\right)$ , then choosing  $k \in \mathbb{N}_0$  and performing steps 1 – 3 of Grover's algorithm leave us with the quantum state

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Measurement will therefore result in a solution with probability

$$\sin^2\left((2k+1)\theta\right) \quad (18)$$

and the optimal  $k \in \mathbb{N}$  is easily computed to be  $\left\lfloor \frac{\pi}{4} \cdot \sqrt{\frac{N}{M}} \right\rfloor$ .

## Grover's quantum search algorithm and query complexity IV

Why does this work?

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Why does this work? Step one creates the *uniform superposition*

$$H^{\otimes n} |0^n\rangle = \frac{1}{2^{\frac{n}{2}}} \sum_{x \in \{0,1\}^n} |x\rangle \quad (19)$$

## Grover's quantum search algorithm and query complexity IV

Why does this work? Step one creates the *uniform superposition*

$$H^{\otimes n} |0^n\rangle = \frac{1}{2^{\frac{n}{2}}} \sum_{x \in \{0,1\}^n} |x\rangle \quad (19)$$

whilst the *Grover iterator* can be rewritten

$$G = D_n S_F, \quad \text{where } D_n = -H^{\otimes n} S_0 H^{\otimes n} \quad (20)$$

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where  $D_n$  is the *diffusion operator on  $n$ -qubits* whose action is

$$\sum_{x \in \{0,1\}^n} \alpha_x |x\rangle \mapsto \sum_{x \in \{0,1\}^n} \left( 2 \cdot \langle \alpha \rangle - \alpha_x \right) |x\rangle \quad (21)$$

where

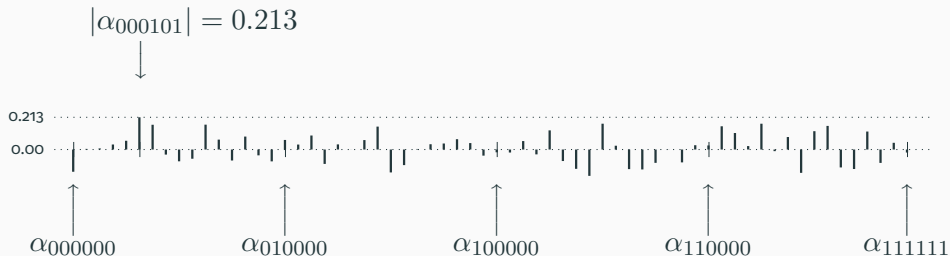
$$\langle \alpha \rangle = \frac{1}{2^n} \sum_{x \in \{0,1\}^n} \alpha_x. \quad (22)$$

# Grover's quantum search algorithm and query complexity V

Let  $N = 2^6$  and  $M = 1$ . Let  $\chi(5) = 1$ .

Fact:  $\frac{\pi}{4} \cdot \sqrt{\frac{2^6}{1}} \approx 6.283$

Step zero: We have a random quantum state  $\sum_{x \in \{0,1\}^n} \alpha_x |x\rangle$ .



Probability of measuring 000101  $\approx 100 \cdot |0.213|^2 = 4.547\%$ .



## Grover's quantum search algorithm and query complexity V

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Step one : Initialise the quantum state to  $|000000\rangle$ .



Probability of measuring 000101  $\approx 100 \cdot |0.000|^2 = 0.000\%$ .

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Let  $N = 2^6$  and  $M = 1$ . Let  $\chi(5) = 1$ .

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Step two : Initialise the uniform superposition by computing  $H^{\otimes n} |0^n\rangle$ .



Probability of measuring 000101  $\approx 100 \cdot |0.125|^2 = 1.563\%$ .

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Step three(1): Apply  $S_F$



Probability of measuring 000101  $\approx 100 \cdot |-0.125|^2 = 1.563\%$ .

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Step three(1): Apply  $S_F$  and then  $D_n$ .



Probability of measuring 000101  $\approx 100 \cdot |0.367|^2 = 13.483\%$ .

## Grover's quantum search algorithm and query complexity V

Let  $N = 2^6$  and  $M = 1$ . Let  $\chi(5) = 1$ .

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Step three(2): Apply  $S_F$



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Step three(2): Apply  $S_F$  and then  $D_n$ .



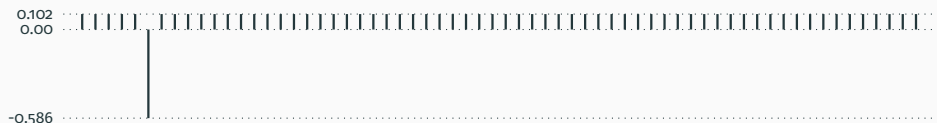
Probability of measuring 000101  $\approx 100 \cdot |0.586|^2 = 34.390\%$ .

## Grover's quantum search algorithm and query complexity V

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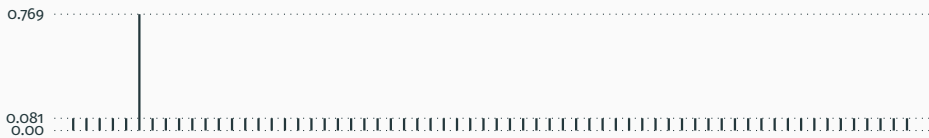
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Step three(3): Apply  $S_F$  and then  $D_n$ .



Probability of measuring 000101  $\approx 100 \cdot |0.769|^2 = 59.138\%$ .

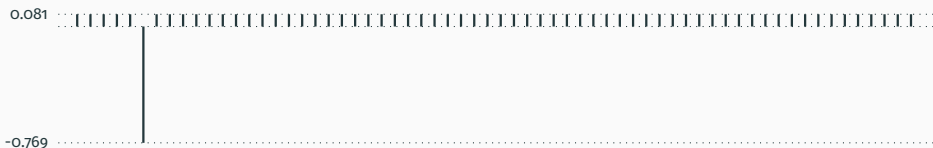


## Grover's quantum search algorithm and query complexity V

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Step three(4): Apply  $S_F$



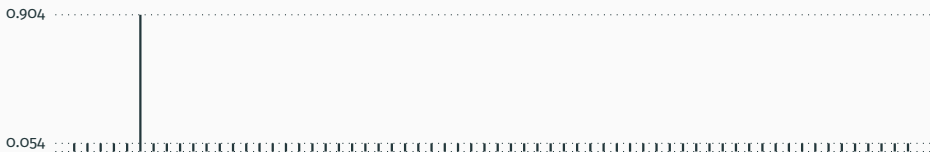
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Step three(4): Apply  $S_F$  and then  $D_n$ .



Probability of measuring 000101  $\approx 100 \cdot |0.904|^2 = 81.638\%$ .

## Grover's quantum search algorithm and query complexity V

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Step three(5): Apply  $S_F$  and then  $D_n$ .



Probability of measuring 000101  $\approx 100 \cdot |0.982|^2 = 96.352\%$ .

## Grover's quantum search algorithm and query complexity V

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Step three(6): Apply  $S_F$



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Probability of measuring 000101  $\approx 100 \cdot |0.998|^2 = 99.659\%$ .

## Grover's quantum search algorithm and query complexity VI

The cost (circuit-depth or circuit-size) of Grover's algorithm, excluding the negligible setup phase of computing  $|\psi_0\rangle = H^{\otimes} |0^n\rangle$  is therefore

$$\left\lfloor \frac{\pi}{4} \cdot \sqrt{\frac{N}{M}} \right\rfloor \cdot \left( \text{Cost}(S_F) + \text{Cost}(D_n) \right). \quad (23)$$

The diffusion step is relatively cheap, costing  $\mathcal{O}(n)$  gates, so the polynomial overhead of Grover's algorithm will usually be dominated by the cost of the quantum oracle.

The number of qubits required is dependent upon the circuit-width of the quantum oracle and is at least  $n$ .

## Grover's quantum search algorithm and query complexity VII

Let  $N = 2^{100}$  and  $M = 1$ .

Say  $\text{Cost}(S_F) + \text{Cost}(D_n) \approx \text{Cost}(S_F) = n^3$

- Query complexity advantage:  $2^{50}$  versus  $2^{100}$ .
- Actual advantage:  $2^{69.93}$  versus  $2^{119.93}$ .
- Each quantum operation will be slower and more expensive.
- Advantageous and easy to run classical search in parallel.
- Disadvantageous to run quantum search in parallel.

$$\left\lfloor \frac{\pi}{4} \cdot \sqrt{\frac{N}{M}} \right\rfloor \cdot \left( \text{Cost}(S_F) + \text{Cost}(D_n) \right). \quad (24)$$

How to optimise?

- Optimise the circuit for  $S_F$  (often involves a tradeoff).



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- Use fewer Grover iterations (lower success probability).
- Tradeoff between information obtained and complexity.

## Quantum circuits and the $\mathcal{MQ}$ evaluation oracle

---

## The $\mathcal{MQ}$ problem again

Minor change in problem:

Assume from now on that we are searching for  $x_1, \dots, x_n$  such that

$$p^{(1)}(x_1, \dots, x_n) = \dots = p^{(m)}(x_1, \dots, x_n) = 1.$$

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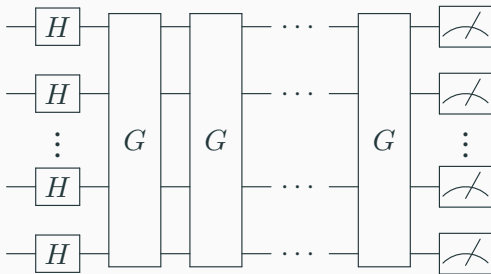
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(Equivalent to the  $\mathcal{MQ}$  problem — simply add +1 to each  $p^{(k)}$ ).

Saves a few gates and makes the complexity slightly more compact.

# The Grover circuit



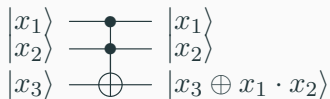
# Building reversible boolean circuits from logical quantum gates



A CNOT gate acting as  $x \oplus y$ .



An X gate acting as negation.



A Toffoli gate acting as  $x \cdot y$ .

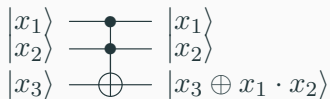
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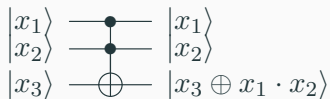
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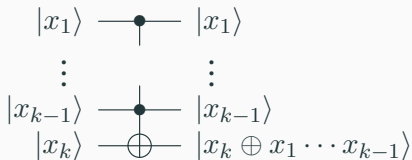
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A Toffoli gate acting as  $x \cdot y$ .



A  $k$ -bit Toffoli gate acting as  $\bigwedge_{i=1}^{k-1} x_i$ .

Classical universal gate set:  $\{\oplus, \neg, \wedge\} \leftrightarrow \{\text{CNOT}, X, \text{Toffoli}\}$

## Constructing the $\mathcal{MQ}$ evaluation oracle

If we have a unitary operator which implements

$$U_E |x\rangle |0^m\rangle \mapsto |x\rangle |p^{(1)}(x)\rangle \dots |p^{(m)}(x)\rangle \quad (25)$$

then we simply exploit the  $|-\rangle = \frac{|0\rangle - |1\rangle}{\sqrt{2}}$  state and use a single  $m + 1$ -bit Toffoli gate  $U_C$  to obtain our quantum oracle as

$$S_F |x\rangle |0^m\rangle |-\rangle = U_E U_C U_E |x\rangle |0^m\rangle |-\rangle \quad (26)$$

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$$= (-1)^{F(x)} |x\rangle |0^m\rangle |-\rangle \quad (29)$$

## Cost of the quantum oracle

$S_F = U_E U_C U_E$ , where

- $U_E$  — evaluates  $p^{(1)}(x), \dots, p^{(m)}(x)$
- $U_C$  — outputs 1 iff  $p^{(1)}(x) = \dots = p^{(m)}(x) = 1$ .

If  $S_F = U_E U_C U_E$ , then the total cost of this operation is therefore

$$2 \times U_E + 1 \times U_C.$$

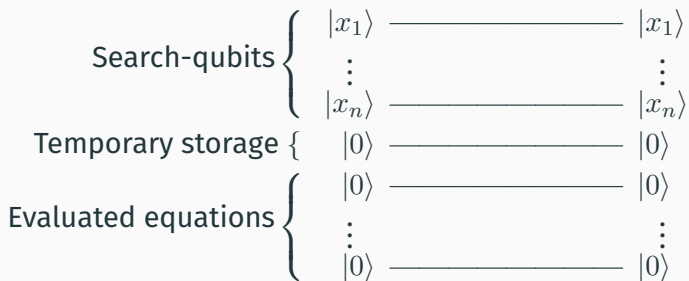
$U_C$  — simply an  $m + 1$ -bit Toffoli gate.

$U_E$  — can be constructed via  $X$ ,  $CNOT$  and Toffoli gate primitives.

# The $\mathcal{MQ}$ quantum oracle

Requirements:  $n + m + 2$  qubits.

- $n$ -qubits — search-space.
- $m$ -qubits — storage for evaluated equations.
- 1-qubit — working memory.
- 1-qubit — for the  $|-\rangle$  state.



An evaluation-based quantum search oracle for the  $\mathcal{MQ}$  problem

# The $\mathcal{MQ}$ quantum oracle I

(Previous work by Schwabe and Westerbaan). Take a single equation

$$p^{(k)}(x_1, \dots, x_n) = \sum_{i=1}^n \sum_{j=i+1}^n a_{i,j}^{(k)} x_i x_j + \sum_{i=1}^n b_i^{(k)} x_i + c^{(k)}. \quad (30)$$

and rewrite it as

$$p^{(k)}(x_1, \dots, x_n) = \sum_{i=1}^n x_i y_i \quad (31)$$

where

$$y_i = b_i^{(k)} + \sum_{j=i+1}^n a_{i,j}^{(k)} x_j. \quad (32)$$



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Evaluation strategy for each equation:

- Compute  $y_i$  onto the workspace via the X and CNOT gates.

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- Compute  $y_i$  onto the workspace via the X and CNOT gates.
- Use a Toffoli gate to add  $x_i y_i$  to the equation register.
- Uncompute  $y_i$  onto the workspace via the X and CNOT gates.

## The $\mathcal{MQ}$ quantum oracle II

$$\begin{aligned} p^{(k)}(x_1, x_2, x_3, x_4) &= x_1x_2 + x_1x_4 + x_2x_3 + x_3x_4 + x_2 + x_4 + 1 \\ &= x_1 \underbrace{(x_2 + x_4)}_{y_1} + x_2 \underbrace{(x_3 + 1)}_{y_2} + x_3 \underbrace{(x_4)}_{y_3} + x_4 \underbrace{(1)}_{y_4} + 1 \end{aligned}$$

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$|x_1\rangle$  \_\_\_\_\_  $|x_1\rangle$

$|x_2\rangle$  \_\_\_\_\_  $|x_2\rangle$

$|x_3\rangle$  \_\_\_\_\_  $|x_3\rangle$

$|x_4\rangle$  \_\_\_\_\_  $|x_4\rangle$

$|0\rangle$  \_\_\_\_\_  $|0\rangle$

$|0\rangle$  \_\_\_\_\_  $|0\rangle$

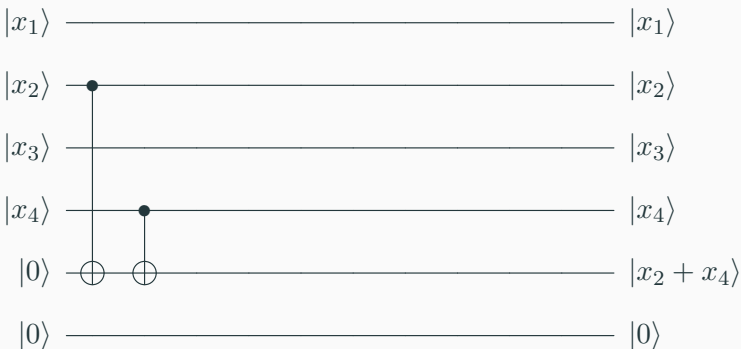
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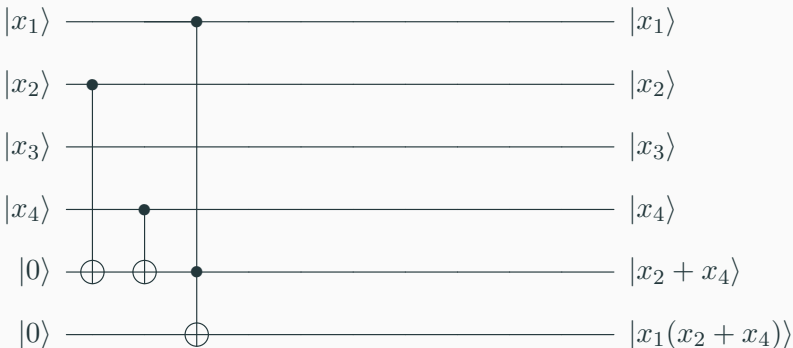
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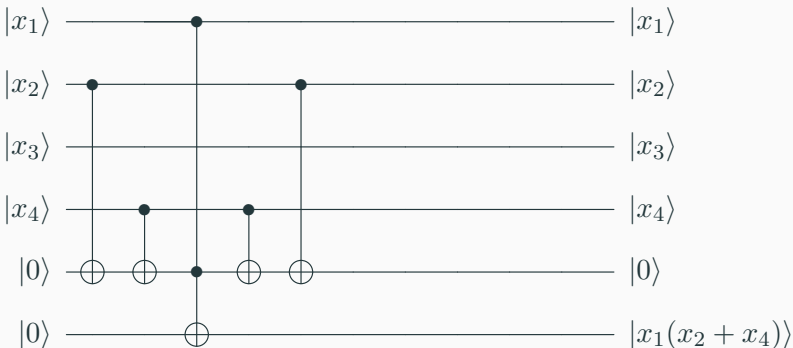
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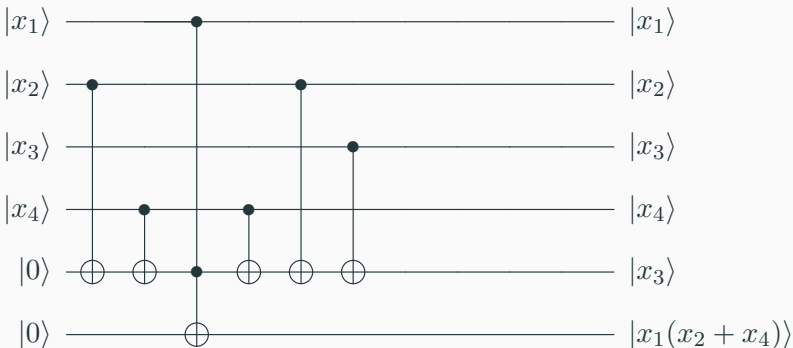
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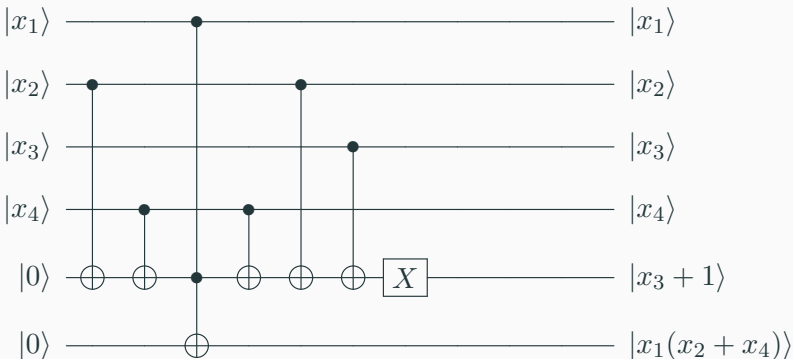
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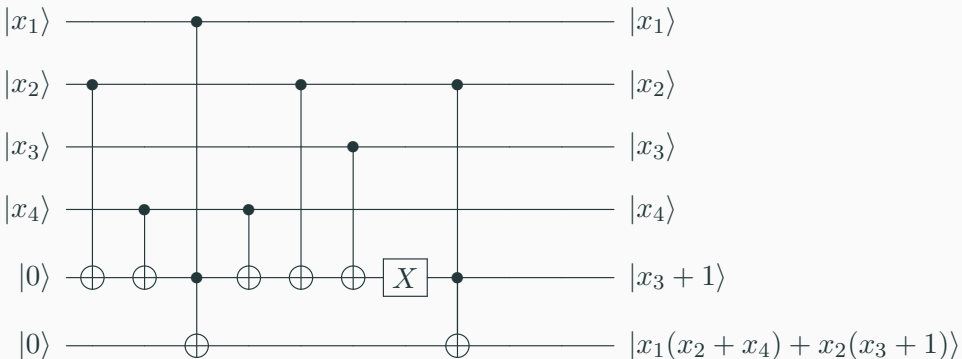
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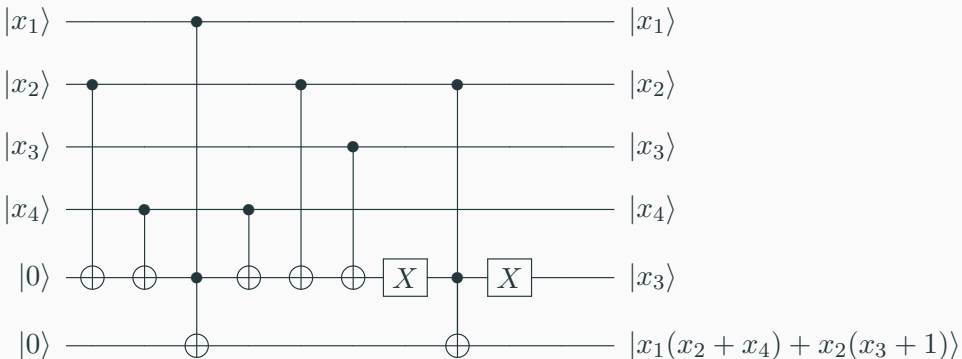
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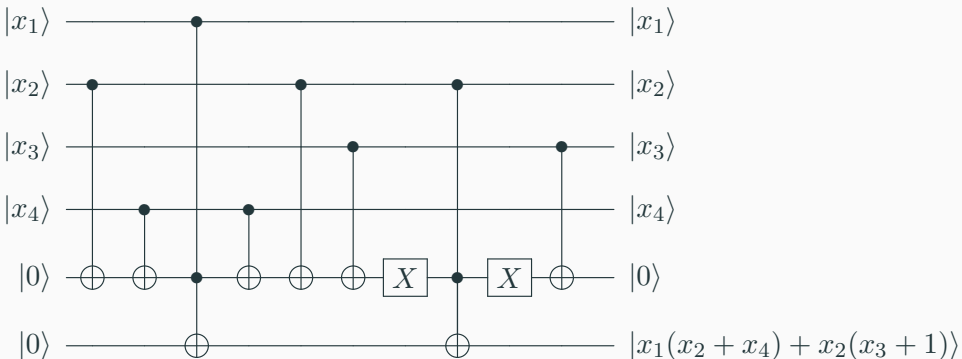
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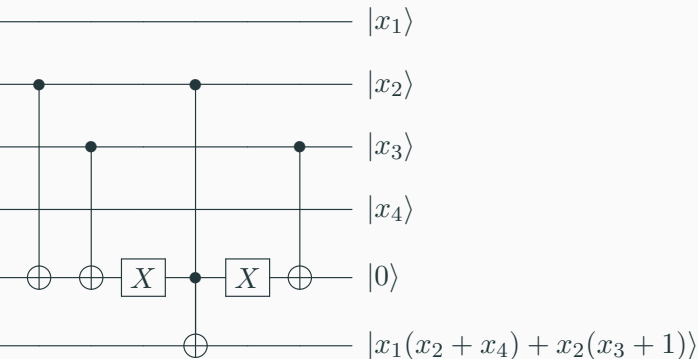
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—  $|x_1\rangle$

—  $|x_2\rangle$

—  $|x_3\rangle$

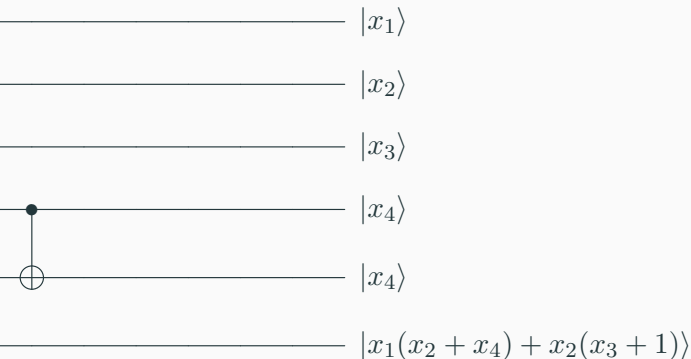
—  $|x_4\rangle$

—  $|0\rangle$

—  $|x_1(x_2 + x_4) + x_2(x_3 + 1)\rangle$

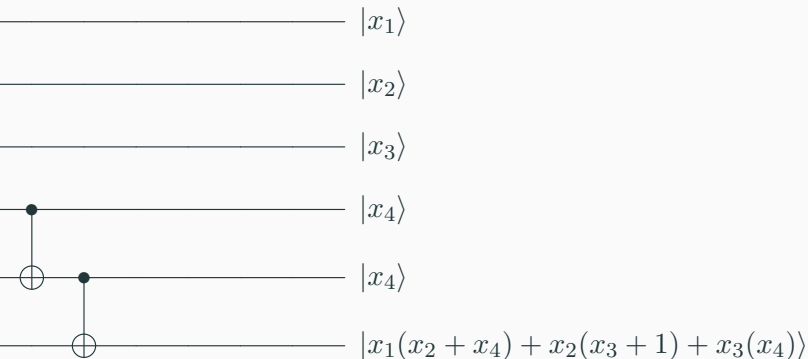
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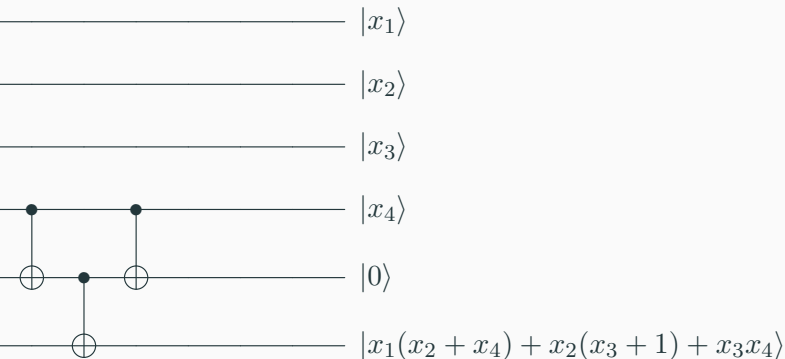
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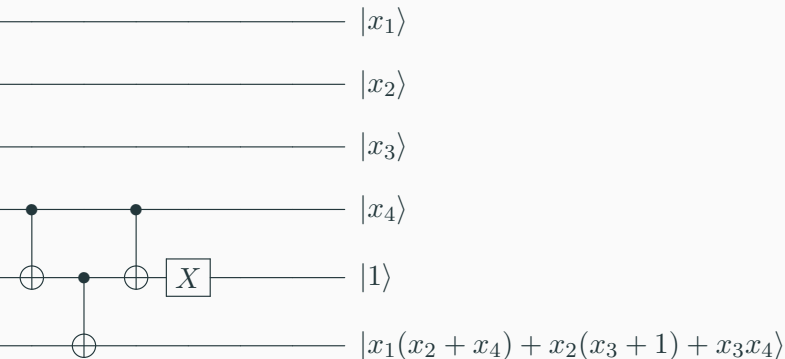
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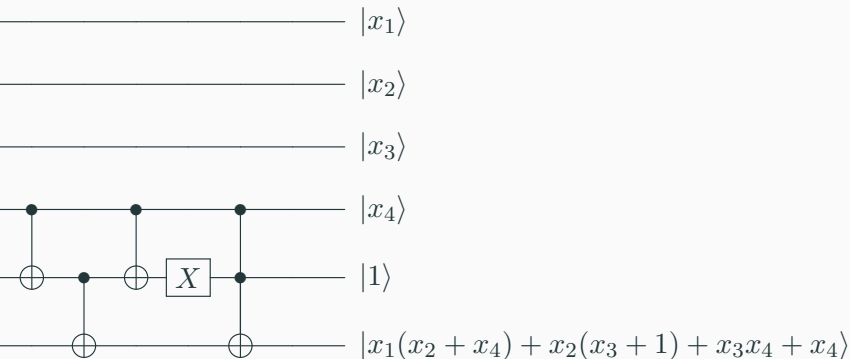
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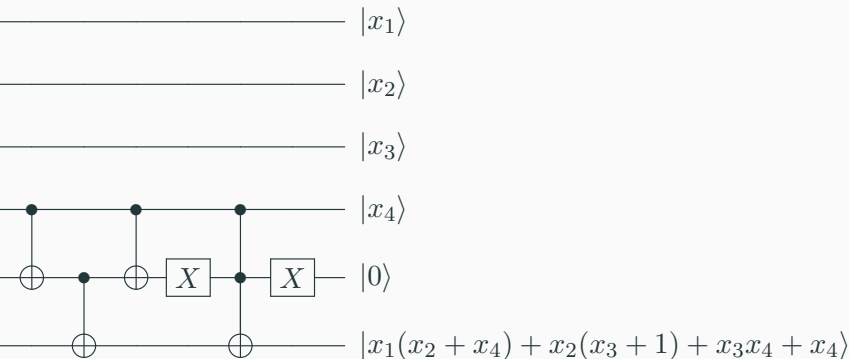
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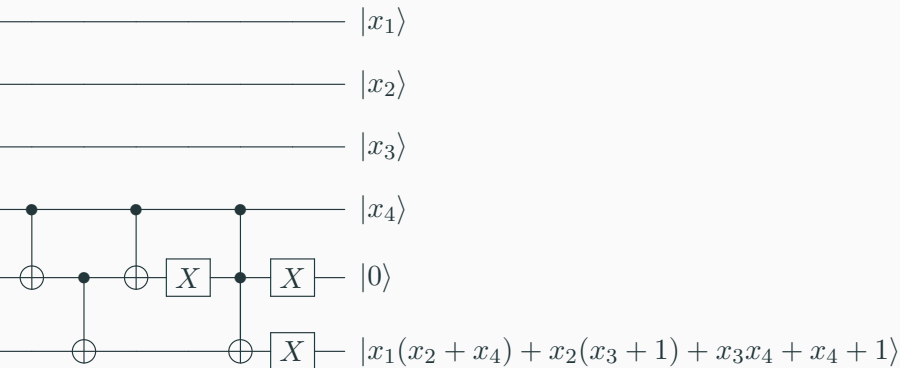
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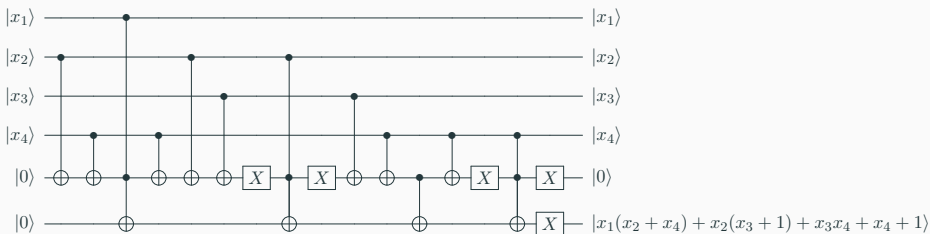
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## The $\mathcal{MQ}$ quantum oracle III

(Previous work by Schwabe and Westerbaan):

$$p^{(k)}(x_1, \dots, x_n) = c^{(k)} + \sum_{i=1}^n x_i y_i \quad \text{where} \quad y_i = b_i^{(k)} + \sum_{j=i+1}^n a_{i,j}^{(k)} x_j \quad (33)$$

Upper bounding of cost to evaluate a single equation:

- $2n + 1$  X gates for the addition of constants (with uncomputation).
- $n^2 - n$  CNOT gates to compute the  $y_i$  (with uncomputation).
- $n$  Toffoli gates adding  $x_i y_i$  to the equation register

Cost to evaluate  $m$  equations and store them in  $m$  registers:

$$m \cdot (n^2 + 2n + 1). \quad (34)$$

We'll ignore specific gate count for now.

## The $\mathcal{MQ}$ quantum oracle IV

(Previous work by Schwabe and Westerbaan)

Cost for  $G$ :  $2 \times$  evaluation cost +  $1 \times$  comparison cost + diffusion cost.

Cost for Grover with  $M = 1$ :

$$\frac{\pi}{4} \cdot 2^{\frac{n}{2}} \cdot \left( 2m(n^2 + 2n + 1) + \text{Toffoli}(m+1) + (\text{Toffoli}(n) + 4n) \right)$$

## **Our adaptation**

---

## Our adaptation I

Aim: optimise the  $\mathcal{MQ}$  oracle so that it breaks the parameters for Gui.

Trick 1: obtain a subset of the bits of the solution, then solve again.

Trick 2: offload computation to preprocessing.

# Our adaptation II

Each  $\mathcal{MQ}$  equation

$$p^{(k)}(x_1, \dots, x_n) = \sum_{i=1}^n \sum_{j=i+1}^n a_{i,j}^{(k)} x_i x_j + \sum_{i=1}^n b_i^{(k)} + c^{(k)}$$

can be written (for a fixed  $0 \leq b \leq n$ )

$$g_1^{(k)}(x_1, \dots, x_{n-b}) + g_2^{(k)}(x_1, \dots, x_n) + g_3^{(k)}(x_{n-b+1}, \dots, x_n)$$

where

$$g_1^{(k)}(x_1, \dots, x_{n-b}) = \sum_{i=1}^{n-b} \sum_{j=i+1}^{n-b} a_{i,j}^{(k)} x_i x_j + \sum_{i=1}^{n-b} b_i^{(k)} x_i$$

$$g_2^{(k)}(x_1, \dots, x_n) = \sum_{i=1}^{n-b} \sum_{j=n-b+1}^n a_{i,j}^{(k)} x_i x_j$$

$$g_3^{(k)}(x_{n-b+1}, \dots, x_n) = \sum_{i=n-b+1}^n \sum_{j=i+1}^n a_{i,j}^{(k)} x_i x_j + \sum_{i=n-b+1}^n b_i^{(k)} x_i + c^{(k)}$$

## Our adaptation III

Trivial that

$$p^{(k)}(\bar{x}_1, \dots, \bar{x}_{n-b}, \bar{x}_{n-b+1}, \dots, \bar{x}_n) = 1$$



$$g_1^{(k)}(\bar{x}_1, \dots, \bar{x}_{n-b}) + g_2^{(k)}(\bar{x}_1, \dots, \bar{x}_{n-b}, \bar{x}_{n-b+1}, \dots, \bar{x}_n) + g_3^{(k)}(\bar{x}_{n-b+1}, \dots, \bar{x}_n) = 1$$

## Our adaptation IV

For a fixed  $0 \leq b \leq n$ , and  $\bar{x}_{n-b+1} \dots \bar{x}_n \in \{0, 1\}^b$

$$g_1^{(k)}(x_1, \dots, x_{n-b}) = \sum_{i=1}^{n-b} \sum_{j=i+1}^{n-b} a_{i,j}^{(k)} x_i x_j + \sum_{i=1}^{n-b} b_i^{(k)} x_i$$

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Let  $i$  be the integer representation of  $\bar{x}_{n-b+1} \dots \bar{x}_n \in \{0, 1\}^b$  and define

$$h_i^{(k)}(x_1, \dots, x_{n-b}) = g_2^{(k)}(x_1, \dots, x_{n-b}, \bar{x}_{n-b+1}, \dots, \bar{x}_n) + g_3^{(k)}(\bar{x}_{n-b+1}, \dots, \bar{x}_n).$$

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For some  $x_1 \dots x_{n-b} \in \{0, 1\}^{n-b}$  and  $0 \leq i < 2^b$  the equation

$$p^{(k)}(x_1, \dots, x_{n-b}) = g_0^{(k)}(x_1, \dots, x_{n-b}) + h_i^{(k)}(x_1, \dots, x_{n-b})$$

is satisfied.

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For some  $0 \leq i < 2^b$  the following equation will be satisfied

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The equation will be satisfied if and only if we have chosen the correct  $x_1, \dots, x_{n-b}$  and the correct  $0 \leq i < 2^b$ .

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  3. For  $i = 1, \dots, 2^b - 1$ :
    - 3.1 Add the difference  $h_i^{(k)}(x_1, \dots, x_{n-b}) - h_{i-1}^{(k)}(x_1, \dots, x_{n-b})$  and test.

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    - 3.1 Add the difference  $h_i^{(k)}(x_1, \dots, x_{n-b}) - h_{i-1}^{(k)}(x_1, \dots, x_{n-b})$  and test.
  4. Restore the equation registers to the state  $g_0^{(k)}(x_1, \dots, x_{n-b})$  via adding  $h_{2^b-1}^{(k)}(x_1, \dots, x_{n-b})$  to the equation registers.

## Our adaptation V

For some  $0 \leq i < 2^b$  the following equation will be satisfied

$$p^{(k)}(x_1, \dots, x_{n-b}) = g_0^{(k)}(x_1, \dots, x_{n-b}) + h_i^{(k)}(x_1, \dots, x_{n-b}). \quad (35)$$

The equation will be satisfied if and only if we have chosen the correct  $x_1, \dots, x_{n-b}$  and the correct  $0 \leq i < 2^b$ . This leads to a strategy:

- Redefine the search problem to be on the variables  $x_1, \dots, x_{n-b}$ .
- Each evaluation of the quantum oracle consists of
  1. Evaluate  $g_0^{(k)}(x_1, \dots, x_{n-b})$  on the equation registers.
  2. Add  $h_0^{(k)}(x_1, \dots, x_{n-b})$  to the equation registers and test if satisfied.
  3. For  $i = 1, \dots, 2^b - 1$ :
    - 3.1 Add the difference  $h_i^{(k)}(x_1, \dots, x_{n-b}) - h_{i-1}^{(k)}(x_1, \dots, x_{n-b})$  and test.
  4. Restore the equation registers to the state  $g_0^{(k)}(x_1, \dots, x_{n-b})$  via adding  $h_{2^b-1}^{(k)}(x_1, \dots, x_{n-b})$  to the equation registers.
  5. Uncompute  $g_0^{(k)}(x_1, \dots, x_{n-b})$ .

## Our adaptation VI

### Remarks

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Total cost

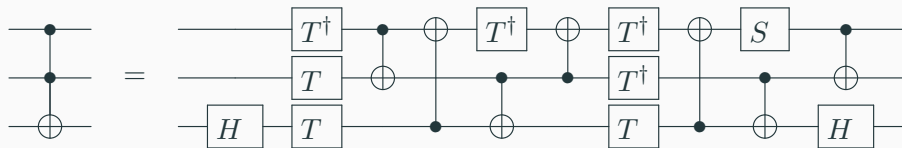
$$\begin{aligned} \frac{\pi}{4} \cdot 2^{\frac{n-b}{2}} \cdot \big( & 2m((n-b)^2 + 2(n-b)) \\ & + 2^b(m(n-b+1) + \text{Toffoli}(m+1)) + m(n-b+1) \\ & + \text{Toffoli}(n) + 4n \big) \end{aligned}$$



## Clifford+T gate costs

Need to choose a quantum gate set to make a fair comparison!

- Clifford+T gate set — advantageous for error correction.
- Clifford+T =  $\{\text{CNOT}, S, S^\dagger, H\} \cup \{T, T^\dagger\}$ .
- Toffoli gates and  $k$ -bit Toffoli gates expensive, but  $\mathcal{O}(k)$  cost.
- We use  $2^b (m + 1)$ -bit Toffoli gates per Grover iteration vs. only 1.
- Can simply convert using best-known implementations.



The logical Toffoli gate decomposed into Clifford+T gates.

# Clifford+T gate costs

Cost in Clifford gates:

$$\frac{\pi}{4} \cdot 2^{\frac{n-b}{2}} \cdot \left( 2m(n-b)^2 + 23m(n-b) + 2^b(m(n-b) + 81m - 160) + 84n - 160 \right)$$

Cost in T gates

$$\frac{\pi}{4} \cdot 2^{\frac{n-b}{2}} \cdot \left( 14m(n-b) + 2^b(52m - 116) + 52(n-b) - 116 \right)$$

Separation of costs is unimportant for cryptanalysis of Gui.

## Impact on key-sizes I

Let's rewind (reverse?) back to Gui.

The security of this system relies upon solving a system of underdefined equations. Essentially there are four parameters

- $n$  — the size of the trapdoor extension field.
- $v$  — the number of vinegar variables (# elements we fix).
- $a$  — the number of minus equations (equations).
- $k$  — the repetition factor (# times we must invert the  $\mathcal{MQ}$  system).

This essentially forms the  $\mathcal{MQ}$  map

$$Gui : \mathbb{F}_2^{n+v} \longrightarrow \mathbb{F}_2^{n-a}, \quad (36)$$

which must be inverted  $k$  times in order to break Gui.

By a few tricks, solving an underdefined ( $m < n$ ) systems of equation can be reduced to that of solving  $m$  equations in  $m$  variables.

## Impact on key-sizes II

Authors initially suggest query complexity to derive parameters.

In a subsequent paper, the same authors suggest a new set of parameters is derived from the new binary  $\mathcal{MQ}$  oracle, using

$$k \cdot 2^{\frac{n-a}{2}} \cdot 2 \cdot (n-a)^3, \quad (37)$$

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## Impact on key-sizes II

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In a subsequent paper, the same authors suggest a new set of parameters is derived from the new binary  $\mathcal{MQ}$  oracle, using

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the number of quantum gates required to break Gui.

- Counts Toffoli/multiple-controlled-Toffoli gates as single gates.
- Clifford+T gate implementation of these primitives is costly.
- Estimation and security holds when all costs are accounted for.

## Impact on key-sizes III

Gui( $n, D, a, v, k$ )

Gui(120, 9, 3, 3, 2): 80-bit security.

Gui(212, 9, 3, 4, 2): 128-bit security.

Gui(464, 9, 7, 8, 2): 256-bit security.

Applying our method and accounting for costs, we obtain

$\lambda$	$n = m$	$k$	$\mathcal{MQ}$ #Gates	Our #Gates	$b$	#Qubits used
80	117	2	$2^{80.99}$	$2^{78.38}$	7	236/229
128	209	2	$2^{129.40}$	$2^{126.26}$	8	420/412
256	457	2	$2^{256.71}$	$2^{252.93}$	10	916/906

Number of Clifford+T gates and qubits required to break Gui.

## Caveats and conclusions

### Caveats:

- Adaptation does NOT break the NIST proposal for Gui, owing to the value MAXDEPTH chosen to be  $2^{64}$ .
- Adaptation cannot be applied to the *low-qubit*  $\mathcal{MQ}$  oracle, which uses half the number of qubits but double the gates.

## Caveats and conclusions

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### Conclusions

- Exploiting structure of problems obviously helps.
- Offload computation to classical preprocessing where possible.
- Hybrid algorithms may provide far better real-world performance.
- Choose security parameters based upon oracle query complexity.



Thanks!

Questions?