

# Quantum Semiprime Factorization: Leveraging Grover's Algorithm for Efficient Prime Decomposition

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## Abstract

*This paper details some of the first steps that our research group has taken towards a practical demonstration of quantum advantage.*

*Grover's search algorithm can be leveraged to efficiently compute the inverse of many functions; this is addressed both from a general perspective, as well as specific applications, including inversion of multiplication, thereby finding the factors of semiprimes with greater efficiency than classical HPC methods and similar existing quantum algorithms.*

*RSA encryption relies on the fact that the multiplication of large prime numbers is a trap-door function- computationally trivial to compute the semiprime product, but classical methods of finding prime components is so computationally expensive that it can be considered nearly impossible at a large scale. This principle is the backbone of modern-day cybersecurity [1].*

**Key terms:** Semiprime, superposition, entanglement, Grover's algorithm, Shor's algorithm, quantum fourier transform, adjoint, RSA encryption, quantum advantage, RSA number, reversible function, trap-door function, quantum circuit, scaling, parallel, CUDA, CUDA-Q, sieve, time complexity.

## 1. Introduction

The practicality of applied quantum computing is a topic of much debate, and there are not yet many examples of quantum computers being able to out-perform their classical counterpart, a feat known as quantum advantage.

Quantum algorithms have been in development at a theoretical level for decades, but the application of these methods is still in early stages. It is quite common to find code which implements a quantum algorithm like Grover's[2] or Shor's[3], but very often these programs are hard-coded, and only work for a small handful of specific values[4].

This paper describes the generalized use of Grover's algorithm as the inverse of some function  $F(x, y) = z$ , with

demonstrated applications inverting the arithmetic functions addition and multiplication. The inverse of multiplication is a solution to prime factorization- a non-trivial problem with significant implications for cybersecurity at a global scope.

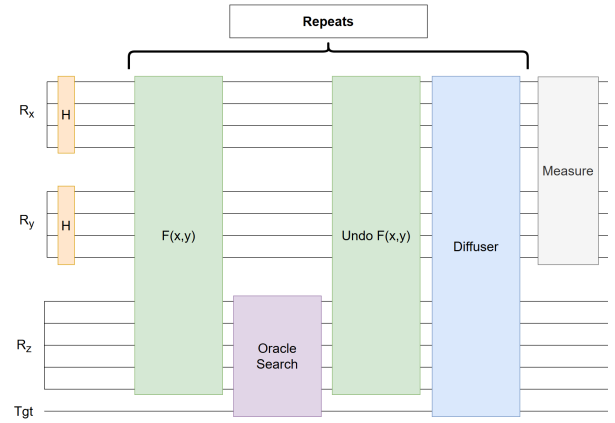


Figure 1. General circuit diagram for Grover's Algorithm, used to find the inputs  $x, y$  for which  $f(x, y) = z$ , for some known output  $z$ .

### 1.1. Goals

The objective of this research is to clearly demonstrate quantum advantage by showing clear speed-up to solve the same problem at the same scale.

While the semiprime factoring problem is just one algorithm where quantum circuits should have significant advantage over parallel factoring, the goal of this research is to concretely show that this particular problem is in fact a clear case of quantum advantage.

One of the major intentions for this paper is demonstration of the broad utility of Grover's algorithm.

Furthermore, the longer-term goal is to identify the types of algorithms that likewise may have significant quantum advantage based upon reversible functions.

## 1.2. Motivation

Semiprime factoring is an ideal problem for demonstrating quantum advantage, as it is both challenging to compute classically, and has wide-reaching impact if efficient solutions are found.

Semiprime factoring is known to have a high time complexity due to the nature of prime number generation with a sieve (filters out all non-primes) and trial division search. This problem in particular becomes complex for large size semiprimes commonly used in encryption, which might be 4096 bits or more. Even a 128-bit semiprime, which is the product of two 64-bit primes requires time complexity that is beyond polynomial, and therefore requires a parallel computer to solve today.

This is important since the simplicity of multiplying two large primes enables fast encryption of data, but the complexity of factoring semiprimes ensures privacy so that data can only be accessed by someone possessing a private key (one of the primes that compose the semiprime used for encryption).

Data privacy hinges on this one-way function, called a trap-door, where factoring is beyond P complexity, but trivial if one possesses the required key.

The only safety now for encryption is scaling the key size (semiprime) to keep it larger than the largest quantum computer number of qubits.

For example, the RSA number 250, which is 892-bits long, was only recently factored using a parallel computer with the best known SP factoring method known right now [5].

The idea of factoring 4096 or bigger is nearly impossible even with the largest parallel computers available [1].

However, a quantum computer has a much lower time complexity if the number of qubits can be scaled to implement a quantum circuit that can factor a semiprime. Ongoing research at China State Key Laboratories suggests that they have been making significant progress in breaking RSA encryption, considering their 48-bit demo using 10 qubits in 2022 [6].

Furthermore, they claim that with the techniques they have been using, it would be possible to break RSA 2048 with a quantum circuit which needs only 372 qubits, far less than the 1,121 qubit machine currently available in the United States with IBM's Condor [7].

If it can be verified that SP factoring with Grover's algorithm uses significantly less qubits than Shor's, has less error, or has less overhead, then the application of quantum semiprime factoring at a meaningful scale will be feasible far sooner.

## 2. Literature Review

### 2.1. Grover's Algorithm

Grover's Algorithm is a quantum computing method that can be used to search a database of  $N$  values in  $O(\sqrt{N})$  time rather than the naive classical time complexity of  $O(N)$ [2].

The exact number of iterations required varies on a case-by-case basis, but is typically expressed as follows: in a search for one matching input state,  $\frac{\pi}{4}\sqrt{N}$  iterations are required, and in a search for  $k$  valid input states,  $\sqrt{\pi 4} \cdot \sqrt{\frac{N}{k}}$  iterations are required, where  $N$  is the size of the search domain.

In most cases,  $N = 2^n$ , where  $n$  is the number of bits needed to represent the target value. This is by no means a steadfast rule, and is meant as a helpful starting point for any confused readers attempting to implement something similar themselves.

Given a function  $f(x) = y$ , where  $x$  is unknown (index, prime factors, sum components, etc.), and  $y$  is known (array value, semiprime/product, sum, etc.), Grover's algorithm effectively takes the role of  $f'(y) = x$ , allowing for a potential speedup in finding whatever input(s) to  $f(x)$  will return  $y$ , provided that it is much faster to compute  $f(x)$  than whatever classical methods might be used to otherwise solve for  $x$  given  $y$ .

This speedup comes from the fact that Grover's algorithm requires  $O(\sqrt{y})$  iterations, each of which have a time complexity proportional to that of  $f(x)$ .

### 2.2. Shor's Algorithm

Shor's algorithm is the most well-known approach to prime factorization in quantum computing. However, there are many drawbacks to this approach, the most notable of which being the need to take measurements mid-circuit and essentially re-run with new values afterwards, which introduces a significant amount of overhead. Even recently improved versions of Shor's algorithm still require a minimum of  $2n + 1$  qubits, and tend to be extremely sensitive to noise, which leads to high error[3, 8].

Furthermore, when it comes to more concrete applications of Shor's algorithm, a full-scale implementation of this algorithm to factor an  $n$ -bit number may require up to  $5n + 1$  qubits for accurate results, and require on the order of  $n^3$  operations [9].

### 2.3. Quantum Factoring Algorithm with Grover Search

S. Whitlock et al. present a method of SP factoring with Grover's algorithm. The implementation in their paper is

highly optimized, and modifies the target value in their circuit from some semiprime  $N$ , to  $M$ , a reduced number uniquely tied to  $N$ , which has unique factors  $p$  and  $q$  which can be used to calculate  $a, b$ , the prime factors of  $N$  [8].

This optimization from  $N$  to  $M$  allows the implementation to ignore the trivial prime factors 2 and 3, and requires fewer qubits than would otherwise be needed to search for prime factors of  $N$ .

While this approach is admirable, it introduces a level of complexity that may hinder a learner’s understanding of the mechanics at work, so the implementation shown in section 3 forgoes this abstraction from  $N$  to  $M$ , and instead implements an oracle which searches directly for prime factors  $a$  and  $b$  of a semiprime  $N$ .

### 3. Methodology

Quantum algorithms take advantage of superposition and entanglement, which enables many methods of computation which are otherwise impossible with a classical computer. For example, by placing a set of  $n$  qubits (otherwise known as a qubit **register**) in superposition, that register simultaneously represents every value which can be represented in  $n$  bits, until measured. Once measured, a register in uniform superposition will collapse into one of these states with equal likelihood.

#### 3.1. The Use of Grover’s Algorithm for General Function Inversion

This use of Grover’s algorithm involves three main parts: some function with one or multiple inputs, the oracle which filters for states in which the desired output was observed, and the diffuser, which increases the probability of observing inputs from the correct state. An overview of this circuit is shown in Figure 1.

The oracle in Grover’s algorithm performs a controlled operation which is meant to differentiate states  $-|x, y\rangle$  that do result in the wanted target  $z$  from states  $|x, y\rangle$  that do not result in the target  $|z\rangle$ .

Figure 2 shows an example of the oracle searching for the state  $z = 15$ , or rather,  $R_z = |001111\rangle$ . To implement a circuit which can selectively target  $z = 15$ , Toffoli gates (otherwise known as NOT, or X gates) are applied on  $R_z[0]$  and  $R_z[1]$ , so that states where  $R_z = |001111\rangle$  will temporarily be in state  $R_z = |111111\rangle$ , and the multi-controlled Toffoli gates [10] will apply only for the desired states. These operations are then performed again, returning all but the target qubit (which has the Z-gate applied) to their original states.

The diffuser in Grover’s algorithm (Figure 3) can then increase the probability of observing  $-|x, y\rangle$  (states which have been marked by the oracle) through a process known as inversion about the mean.

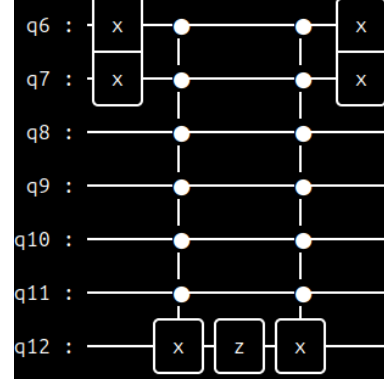


Figure 2. Example circuit diagram for Grover’s oracle marking entangled states of the output register ( $R_z = [q_0, q_{11}]$ ) via the target qubit ( $q_{12}$ ), where  $R_z = |15\rangle$ .

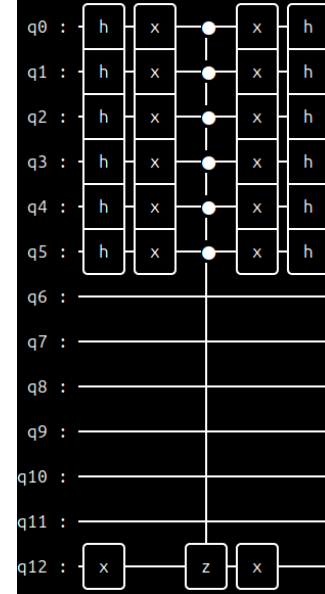


Figure 3. Example circuit diagram for a diffuser which amplifies the probability that measurement will result in a state which has been marked by the oracle. Input registers  $R_x$  and  $R_y$  ( $[q_0, q_2]$  and  $[q_3, q_5]$ ) are now more likely to be measured in a state such that  $f(R_x, R_y)$  results in  $R_z = |z\rangle$ .

#### 3.2. Semiclassical Arithmetic

As a proof of concept (prior to the release of [8]), a function  $f(x, y) = x + y$  is defined such that  $R_x = |x\rangle$ ,  $R_y = |y\rangle$ , and after performing  $f(R_x, R_y)$ , the output register  $R_z$  shall be in the state  $|x + y\rangle$ .

This approach was implemented in a similar manner to a classical full-adder, and has been observed to work in simulation with perfect accuracy [11].

The semiclassical full-adder indeed works with Grover’s

algorithm. Input states were filtered to only measure inputs which resulted in a sum matching some given value. This is a trivial problem however, and only has potential usefulness due to the relationship between multiplication and addition.

Unfortunately, semiclassical multiplication requires direct measurement of  $R_x$  or  $R_y$ , which would collapse the superposition, so a different approach was needed.

### 3.3. Arithmetic in the Quantum Fourier Domain

The quantum fourier transform, QFT, is a very useful operation, typically used to transform a qubit register from the computational basis (binary) to the fourier basis (phase). The QFT operation is defined in great detail by Perez et al. [12].

To summarize, given a qubit register  $R_x = |x\rangle$  with  $n$  qubits in the computational basis, there are  $d = 2^n$  potential states  $|k\rangle$  in  $|x\rangle = \{|0\rangle, |1\rangle, \dots, |d-1\rangle\}$ , the QFT operation in this basis would be defined as:

$$QFT|x\rangle = \frac{1}{\sqrt{d}} \sum_{k=0}^{d-1} e^{i\frac{2\pi xk}{d}} |k\rangle \quad (1)$$

\*It is worth noting that in some cases, a QFT on some  $R_x$  may require a series of SWAP operations  $\{SWP(q_0, q_{n-1}), SWP(q_1, q_{n-2}), \dots, SWP(q_{\frac{n}{2}-1}, q_{\frac{n}{2}})\}$  in order for future arithmetic to work as expected.

Once in the fourier basis, a specific number  $x$  may be encoded by applying some rotation  $\omega_k$  to qubits  $q_k \in R_x |q_k = R_x[k]$ . The rotation  $\omega_k$  applied to each qubit  $q_k$  can be defined as  $\omega_k = \omega^{xk} = e^{i\frac{2\pi xk}{d}}$ .

The inverse QFT operation, IQFT is used to return a qubit register  $R_x$  from the fourier basis to the computational basis, retaining the values which have been mapped and operations which have been applied to  $R_x$ .

Addition of two qubit registers  $R_a$  in state  $|a\rangle$  and  $R_b$  in state  $|b\rangle$  in the fourier basis ( $\phi ADD(|a\rangle, |b\rangle) = |a+b\rangle$ ) may be performed by applying controlled rotation gates  $R_k$  which rotate qubits  $b_j \in R_b$  by  $\omega_j = e^{\frac{2\pi i}{2^j}}$  about the  $z$ -axis for each qubit  $a_l |l \geq k, j = 1 + l - k$  [12].

Multiplication in the quantum fourier domain can be done similarly. In terms of a quantum circuit generation implementation, simply treat the previously defined addition method as a function which adds the value of one register to another, and add some constant multiple  $C$  as a parameter, adjusting  $\omega_j$  to  $\omega_{j,C} = e^{C \cdot \frac{2\pi i}{2^j}}$ . Now, to send the product of the two values in  $R_x$  and  $R_y$  to  $R_z$ , do as follows: Add  $|R_y|$  controlled  $\phi ADD(R, q_c, C)$  gates to each  $q_x$  in  $R_x$ , which will add the value of  $|C \cdot x\rangle$  to  $R_z$  where  $C = 2^u$  for each  $q_u \in R_y \iff q_y = |1\rangle$ . This is, in essence, performing repeated scaled addition from  $R_x$

onto  $R_z$ , but the scaling factor is the powers of two corresponding to qubits in  $R_y$ , which in tandem is cumulatively equivalent to adding  $|x \cdot y\rangle$  to  $R_z$  [12].

### 3.4. Semiprime Factoring with Grover's Algorithm

After implementing the QFT arithmetic methods and the oracle and diffuser for grover's, it is trivial to implement a SP factoring circuit. Our implementation follows the diagram in Figure 1, while substituting  $F(x, y)$  for the  $\phi MULT(R_x, R_y)$  operation described at the end of section 3.3.

The only fine-tuning necessary here was to add additional bits to the  $R_z$  register. Typically, only  $\lceil \log_2(N) \rceil$  qubits are needed to represent some number  $N$ , and while that is true, we experienced something that might be considered quantum integer overflow before increasing the number of qubits in  $R_z$ . Prime factors  $a, b$  of  $N$  may be as little as one bit smaller than  $N$ , and due to the fact that  $R_x$  and  $R_y$  representing  $a, b$  are in superposition, that meant that all numbers up to  $2^{n-1}$  were being included in the superposition for both  $a$  and  $b$ , resulting in products far larger than could be represented in  $R_z$ , which in turn led to false positives due to products that were some sum of powers of 2 greater than  $n$ .

The temporary solution we found was to simply decrease the size of  $R_x$  and  $R_y$  to  $\lceil \log_2(\frac{N}{3}) \rceil$ , and set the size of  $R_z$  to double that, to ensure no further overflow could occur.

## 4. Results and Discussion

To date, we have preliminary results based on CUDA-Q simulation of quantum circuits, which have been compared to classic parallel methods to generate primes with a sieve and search them for modulo zero factoring.

This work is still in progress, but we have a working implementation of Grover's algorithm with the CUDA-Q simulation which can estimate the time to factor any semiprime up to the current scaling limits of the CUDA-Q simulation[13].

We first plan to compare the Grover quantum circuit to our best parallel sieve and modulo search with this scale-out with SDSC Expanse.

### 4.1. Accuracy and Limitations

On our CSU Chico A100 system, we have been able to scale this implementation up to a total of 32 qubits, used to find the prime factors  $47 \times 13 = 611$ , where 611 is a 10-bit semiprime. This performance is due to the suboptimality of the current implementation, as we have not

yet introduced the search-term reduction from  $N$  to  $M$  as is done by S. Whitlock et al.[8].

Due to the exploratory nature of this work, the initially sub-optimal nature of our implementation is intentional, because it allows for a more direct view of the inputs and outputs in our circuit.

For context, our current implementation uses  $4\lceil\log_2(\frac{N}{3})\rceil$  qubits to find the prime factors of an  $n$ -bit semiprime, whereas the design by S. Whitlock et al.[8] uses up to  $2n - 5$  qubits, which allowed them to simulate circuits which found prime factors of up to 35-bit semiprimes, using up to a total of 65 qubits. The limiting factor for our simulations is VRAM, so it is expected that after implementing the optimizations from S. Whitlock et al., our simulations should be able to scale up to 18-bit semiprimes on the current equipment.

## 5. Conclusion

Grover’s algorithm has been demonstrated to be a highly versatile function, able to invert almost any function that can be implemented on a quantum computer with inputs in superposition.

Quantum methods of SP factoring have been developing at a quick pace, and it will most likely not be very long before practical examples of quantum advantage may be seen.

## 6. Future Work

The current priority regarding the quantum circuit is to implement the optimizations mentioned by S. Whitlock et al. to substitute the search term  $N$  with the smaller search term  $M$ , which should reduce the number of qubits needed from  $4\lceil\log_2(\frac{N}{3})\rceil$  to  $2n - 5$ .

We have recently gained access to resources at the San Diego Supercomputing Center (SDSC) Expanse cluster via an NSF allocation grant. This equipment will allow us to simulate on far larger problem scales, and by tracking the number of operations (quantum gates) used at different problem scales and comparing those counts against the number of operations performed in our classical parallel solution, we should be able to simulate circuits at a scale much larger than our current limitation of  $\approx 32$  qubits.

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