# QUANTUM COMPUTATION AND QUANTUM INFORMATION: THE QUANTUM FOURIER TRANSFORM

1.

We consider the linear map in  $\mathbb{C}^N$  which acts on the computational basis as

$$|j\rangle\mapsto \frac{1}{\sqrt{N}}\sum_{k=0}^{N-1}e^{\frac{2i\pi jk}{N}}\,|k\rangle$$

Let A be the matrix of the transformation in the computational basis.

$$\forall (k,l) \in [0,N-1]^2, \quad a_{kl} = \frac{1}{\sqrt{N}} e^{\frac{2i\pi kl}{N}}$$

The adjoint matrix  $A^{\dagger}$  is then

$$\forall (k,l) \in [0, N-1]^2, \quad b_{kl} = a_{lk}^* = \frac{1}{\sqrt{N}} e^{-\frac{2i\pi kl}{N}}$$

We compute the coefficient k, l of the product  $AA^{\dagger}$ :

$$\begin{split} \forall (k,l) \in [\![0,N-1]\!]^2, \quad c_{kl} &= \sum_{j=0}^{N-1} a_{kj} b_{jl} \\ &= \frac{1}{N} \sum_{j=0}^{N-1} e^{\frac{2i\pi j}{N}(k-l)} \\ &= \frac{1}{N} \sum_{j=0}^{N-1} (e^{\frac{2i\pi}{N}(k-l)})^j \\ &= \begin{cases} \frac{1}{N} \frac{1 - (e^{\frac{2i\pi}{N}(k-l)})^N}{1 - e^{\frac{2i\pi}{N}(k-l)}} = 0 & \text{if } e^{\frac{2i\pi}{N}(k-l)} \neq 1, \\ 1 & \text{if } e^{\frac{2i\pi}{N}(k-l)} = 1. \end{cases} \\ &= \begin{cases} 0 & \text{if } k \neq l, \\ 1 & \text{if } k = l. \end{cases} \\ &= \delta_{kl} \end{split}$$

which shows that  $AA^{\dagger} = A^{\dagger}A = I$  i.e. A is unitary.

2.

Here the dimension of the state space is  $N=2^n$ . The Fourier transform of the n qubit state  $|00...0\rangle$  is

$$A|0\rangle = \frac{1}{\sqrt{N}} \sum_{k=0}^{N-1} |k\rangle$$

we can write k in binary  $k_{n-1} \dots k_1 k_0$ 

$$A|0\rangle = \frac{1}{2^{n/2}} \sum_{k_0, k_1, \dots, k_0} |k_{n-1} \dots k_1 k_0\rangle$$

or in product representation,

$$= \frac{1}{2^{n/2}} \underbrace{(|0\rangle + |1\rangle)(|0\rangle + |1\rangle) \dots (|0\rangle + |1\rangle)}_{\substack{n \text{ qubits}}}$$

Let  $N=2^n$  and  $Y=(y_k)_{k\in [0,N-1]}$  be the classical fourier transform of  $X=(x_k)_{k\in [0,N-1]}$ .

$$\forall k \in [0, N-1], \quad y_k = \sum_{j=0}^{N-1} e^{\frac{2i\pi kj}{2^n}} x_j$$

The factor  $\frac{1}{\sqrt{N}}$  is omitted for clarity. We can write j in binary  $j_{n-1} \dots j_1 j_0$ 

$$\begin{aligned} y_k &= \sum_{j_0,j_1,\dots,j_{n-1}=0}^1 e^{\frac{2i\pi k(2^{n-1}j_{n-1}+\dots+2j_1)}{2^n}} x_j \\ &= \sum_{j_1,\dots,j_{n-1}=0}^1 e^{\frac{2i\pi k(2^{n-1}j_{n-1}+\dots+2j_1)}{2^n}} x_{j_{n-1}\dots j_10} + \sum_{j_1,\dots,j_{n-1}=0}^1 e^{\frac{2i\pi k(2^{n-1}j_{n-1}+\dots+2j_1+1)}{2^n}} x_{j_{n-1}\dots j_11} \\ &= \sum_{j_1,\dots,j_{n-1}=0}^1 e^{\frac{2i\pi k(2^{n-1}j_{n-1}+\dots+2j_1)}{2^n}} x_{j_{n-1}\dots j_10} + e^{\frac{2i\pi k}{2^n}} \sum_{j_1,\dots,j_{n-1}=0}^1 e^{\frac{2i\pi k(2^{n-1}j_{n-1}+\dots+2j_1)}{2^n}} x_{j_{n-1}\dots j_11} \\ &= \sum_{j_1,\dots,j_{n-1}=0}^1 e^{\frac{2i\pi k(2^{n-2}j_{n-1}+\dots+j_1)}{2^{n-1}}} x_{j_{n-1}\dots j_10} + e^{\frac{2i\pi k}{2^n}} \sum_{j_1,\dots,j_{n-1}=0}^1 e^{\frac{2i\pi k(2^{n-2}j_{n-1}+\dots+j_1)}{2^{n-1}}} x_{j_{n-1}\dots j_11} \end{aligned}$$

We see the first sum is the  $k^{th}$  coefficient of the FT of the sequence  $(x_{2k})_{k \in [0,N/2-1]}$  and the second is the  $k^{th}$  coefficient of the FT of  $(x_{2k+1})_{k \in [0,N/2-1]}$ . This shows that to compute FT of sequence of length N, we have to compute 2 FT of sequence of length  $\frac{N}{2}$  and do 2N complex additions/multiplications. The complexity of the operation T(N) follows the recurrence:

$$T(N) = 2T(\frac{N}{2}) + 2N$$

We can use the Master theorem [1]:

**Theorem.** Let  $a \ge 1$  and b > 1 be constants, let f(n) be a function, and let T(n) be defined on the non negative integers by the recurrence

$$T(n) = aT(\frac{n}{h}) + f(n)$$

where we interpret  $\frac{n}{h}$  to mean either  $\lfloor \frac{n}{h} \rfloor$  or  $\lceil \frac{n}{h} \rceil$ . Then T(n) has the following asymptotic bounds:

- (1) If  $f(n) = O(n^{\log_b a \epsilon})$  for some constant  $\epsilon > 0$ , then  $T(n) = \Theta(n^{\log_b a})$ .
- (2) If  $f(n) = \Theta(n^{\log_b a})$ , then  $T(n) = \Theta(n^{\log_b a} \log n)$ .
- (3) If  $f(n) = \Omega(n^{\log_b a + \epsilon})$  for some constant  $\epsilon > 0$ , and if  $af(\frac{n}{b}) \leqslant cf(n)$  for some constant c < 1 and n sufficiently large, then  $T(n) = \Theta(f(n))$ .

Here we are in the second case of the theorem, so  $T(N) = \Theta(N \log(N)) = \Theta(n2^n)$ .

Instead of  $\mathbb{C}$ , the Fourier transform may be used in any ring as soon as we are given a Nth root of unity. The book The design and analysis of computer algorithms [2] provides an overview of the FFT, an algorithm using bits operations and application to fast integer multiplication.

**5.** 

The inverse Fourier Transform

$$|j\rangle\mapsto \frac{1}{\sqrt{N}}\sum_{k=0}^{N-1}e^{-\frac{2i\pi jk}{N}}|k\rangle$$

is the adjoint of the Fourier Transform. The quantum circuit of figure 1 is obtained from the FT's circuit, replacing each  $R_k$  gate by its adjoint

$$R_k^{\dagger} = \begin{bmatrix} 1 & 0\\ 0 & e^{-\frac{2i\pi}{2^k}} \end{bmatrix}$$



FIGURE 1. Quantum circuit for IFT.



Figure 2. Sequence of controlled U.

In figure 2, the t qubits of the first register are prepared with  $|j\rangle = |j_{t-1}\dots j_1j_0\rangle$ , the second register is prepared with some state  $|u\rangle$ . After the first controlled-U operation, the state is  $|j\rangle |U^{j_02^0}u\rangle$ . After the second controlled-U, the state is  $|j\rangle |U^{j_12^1}U^{j_02^0}u\rangle = |j\rangle |U^{j_02^0+j_12^1}u\rangle$  and so on. The final state is  $|j\rangle |U^{j_02^0+j_12^1+\dots+j_{t-1}2^{t-1}}u\rangle = |j\rangle |U^ju\rangle$ .

## 8.

By linearity, the phase estimation algorithm takes input  $|0\rangle |\Sigma_{u\in A}c_u|u\rangle$ , where A is some orthonormal basis of eigenstates of U, to output  $\sum_{u\in A}c_u|\widetilde{\varphi_u}\rangle |u\rangle$ , where  $\widetilde{\varphi_u}$  is an estimation of the phase of the eigenvalue associated with eigenstate u. If we fix  $u_0\in A$  beforehand, the probability to measure  $\widetilde{\varphi_{u_0}}$  when measuring the first register in the computational basis is

$$\begin{split} (\sum_{u \in A} c_u^* \, \langle \widetilde{\varphi_u} | \, \langle u |) P_{\widetilde{\varphi_{u_0}}} \otimes I(\sum_{u \in A} c_u \, | \widetilde{\varphi_u} \rangle \, | u \rangle) &= (\sum_{u \in A} c_u^* \, \langle \widetilde{\varphi_u} | \, \langle u |) (\sum_{\substack{u \in A \\ \widetilde{\varphi_u} = \widetilde{\varphi_{u_0}} \rangle}} c_u \, | \widetilde{\varphi_u} \rangle \, | u \rangle) \\ &= (\sum_{u \in A} c_u^* \, \langle \widetilde{\varphi_u} | \, \langle u |) (\sum_{\substack{u \in A \\ \widetilde{\varphi_u} = \widetilde{\varphi_{u_0}} \rangle}} c_u \, | \widetilde{\varphi_{u_0}} \rangle \, | u \rangle) \\ &= \sum_{\substack{v \in A \\ u \in A \\ \widetilde{\varphi_u} = \widetilde{\varphi_{u_0}} \rangle}} c_v^* c_u \, \langle \widetilde{\varphi_v} | \widetilde{\varphi_u} \rangle \, \langle v | u \rangle \\ &= \sum_{\substack{u \in A \\ \widetilde{\varphi_u} = \widetilde{\varphi_{u_0}} \rangle}} |c_u|^2 \\ &\geqslant |c_{u_0}|^2 \end{split}$$



FIGURE 3. Phase estimation circuit with t = 1.

I is the identity operator of whatever state space U operates on, while  $P_{\widetilde{\varphi_{u_0}}}$  is the orthonormal projector onto the space generated by the vector  $|\widetilde{\varphi_{u_0}}\rangle$  of the computational basis. Besides, following the analysis of the book,  $\widetilde{\varphi_{u_0}}$  is an approximation to  $\varphi_{u_0}$  to an accuracy  $2^{-n}$  with probability at least  $1 - \epsilon$  if we make use of  $t = n + \lceil \log(2 + \frac{1}{2\epsilon}) \rceil$  bits in the first register. We conclude we get the desired approximation of  $\varphi_{u_0}$  at the end of the phase estimation algorithm with probability at least  $|c_{u_0}|^2(1 - \epsilon)$ .

9.

U being unitary with eigenvalues -1 and +1, the state space is the direct sum of the two orthogonal eigenspaces  $E_{-1} \oplus E_1$ . Thus we can uniquely decompose any  $|\psi\rangle = |\psi_{-1}\rangle + |\psi_{+1}\rangle$ , with  $|\psi_{-1}\rangle \in E_{-1}$  and  $|\psi_{+1}\rangle \in E_{+1}$ . Then  $-1 = e^{i\pi} = e^{2i\pi 0.1}$  and  $1 = e^0 = e^{2i\pi 0.0}$  shows that is sufficient to make use of t = 1 wire in the first register in the phase estimation procedure to read directly the phase of any eigenvector. If we use  $|0\rangle |\psi\rangle$  as input in the circuit of figure 3, the output before the final measurement will be  $|0\rangle |\psi_{+1}\rangle + |1\rangle |\psi_{-1}\rangle$ . When we measure the first register, we obtain 0 with probability

$$(\langle 0 | \langle \psi_{+1} | + \langle 1 | \langle \psi_{-1} | ) P_0 \otimes I(|0\rangle | \psi_{+1} \rangle + |1\rangle | \psi_{-1} \rangle) = (\langle 0 | \langle \psi_{+1} | + \langle 1 | \langle \psi_{-1} | ) (|0\rangle | \psi_{+1} \rangle)$$

$$= \langle 0 | 0 \rangle \langle \psi_{+1} | \psi_{+1} \rangle$$

$$= \langle \psi_{+1} | \psi_{+1} \rangle$$

or 1 with probability

$$(\langle 0 | \langle \psi_{+1} | + \langle 1 | \langle \psi_{-1} | ) P_1 \otimes I(|0\rangle | \psi_{+1} \rangle + |1\rangle | \psi_{-1} \rangle) = (\langle 0 | \langle \psi_{+1} | + \langle 1 | \langle \psi_{-1} | ) (|1\rangle | \psi_{-1} \rangle)$$

$$= \langle 1 | 1 \rangle \langle \psi_{-1} | \psi_{-1} \rangle$$

$$= \langle \psi_{-1} | \psi_{-1} \rangle$$

The state will collapse respectively into  $\frac{1}{\sqrt{\langle \psi_{+1}|\psi_{+1}\rangle}} |0\rangle |\psi_{+1}\rangle$  or  $\frac{1}{\sqrt{\langle \psi_{-1}|\psi_{-1}\rangle}} |1\rangle |\psi_{-1}\rangle$ . Thus if we read 0 in the first register, that means that we have an eigenvector associated to eigenvalue +1 in the second register, and if we read 1 in the first register, that means that we have an eigenvector associated to eigenvalue -1 in the second register.

Once we have noticed that the FT in dimension  $N = 2^1$  is just the Hadamard operator, we conclude the phase estimation circuit in this particular case is the just the same as the circuit of exercice 4.34.

10.

$$x^{2} = 25 = 4$$
  
 $x^{3} = 20 = -1$   
 $x^{4} = 4^{2} = 16$   
 $x^{5} = 16 \times 5 = 80$   
 $= 17$   
 $x^{6} = (-1)^{2} = 1$ 

11.

**Theorem** (Euler). For  $N \in \mathbb{N}^*$ , let

$$\varphi(N)=\#\{m\in[\![1,N]\!],m\wedge N=1\}$$

We have

$$\forall x \in \mathbb{N}^*, \quad x \land N = 1 \Rightarrow x^{\varphi(N)} = 1 \mod N$$

Then by definition of the order  $r, r \leq \varphi(N) \leq N$ .

Since  $x \wedge N = 1$ , from Bezout's Theorem  $\exists (u, v) \in \mathbb{Z}^2$  such that ux + vN = 1 that is  $\exists u$  such that ux = 1 mod N which shows that x has a multiplicative inverse  $x^{-1} = u$  in the ring  $(\frac{\mathbb{Z}}{N\mathbb{Z}}, +, \times)$ . We define the linear map U' on  $(\mathbb{C}^2)^{\otimes L} \cong \mathbb{C}^{2^L}$  that acts on the computational basis as

$$\forall y \in \{0,1\}^L, \quad U' \left| y \right\rangle = \left\{ \begin{array}{ll} \left| x^{-1}y \mod N \right\rangle & \text{if } y < N, \\ y & \text{if } y \in [\![N,2^L-1]\!]. \end{array} \right.$$

We have

$$\forall y_1, y_2 \in \{0, 1\}^L, \quad \langle y_1 | U(y_2) \rangle = 1 \Leftrightarrow y_1 = y_2 \in \llbracket N, 2^L - 1 \rrbracket \text{ or } (y_1, y_2 < N \text{ and } xy_2 = y_1 \mod N)$$

$$\Leftrightarrow y_1 = y_2 \in \llbracket N, 2^L - 1 \rrbracket \text{ or } (y_1, y_2 < N \text{ and } \exists k \in \mathbb{Z}, xy_2 = y_1 + kN)$$

$$\Leftrightarrow y_1 = y_2 \in \llbracket N, 2^L - 1 \rrbracket \text{ or } (y_1, y_2 < N \text{ and } \exists k \in \mathbb{Z}, y_2 = x^{-1}y_1 + x^{-1}kN)$$

$$\Leftrightarrow y_1 = y_2 \in \llbracket N, 2^L - 1 \rrbracket \text{ or } (y_1, y_2 < N \text{ and } \exists k' \in \mathbb{Z}, y_2 = x^{-1}y_1 + k'N)$$

$$\Leftrightarrow y_1 = y_2 \in \llbracket N, 2^L - 1 \rrbracket \text{ or } (y_1, y_2 < N \text{ and } x^{-1}y_1 = y_2 \mod N)$$

$$\Leftrightarrow \langle U'(y_1) | y_2 \rangle = 1$$

so, since  $\langle U'(y_1)|y_2\rangle$ ,  $\langle y_1|U(y_2)\rangle \in \{0,1\}$ ,

$$\forall y_1, y_2 \in \{0, 1\}^L, \quad \langle y_1 | U(y_2) \rangle = \langle U'(y_1) | y_2 \rangle$$

This shows that  $U' = U^{\dagger}$ , since it is obvious that U is invertible and  $U^{\dagger} = U^{-1}$ , we have shown that U is unitary.

13.

 $(|u_s\rangle)_{s\in \llbracket 0,r-1\rrbracket}$  is defined to be the IFT of the sequence  $(|x^k\mod N\rangle)_{k\in \llbracket 0,r-1\rrbracket}$ :

$$\forall s \in [0, r-1], \quad |u_s\rangle = \frac{1}{\sqrt{r}} \sum_{k=0}^{r-1} e^{-\frac{2i\pi sk}{r}} |x^k \mod N\rangle$$

Thus the equalities

$$\forall k \in [0, r-1], \quad |x^k \mod N\rangle = \frac{1}{\sqrt{r}} \sum_{s=0}^{r-1} e^{\frac{2i\pi sk}{r}} |u_s\rangle$$

just state the fact that  $(|x^k \mod N\rangle)_{k \in [0,r-1]}$  is the FT of the sequence  $(|u_s\rangle)_{s \in [0,r-1]}$ . Let's check this. Let  $k \in [0,r-1]$ ,

$$\frac{1}{\sqrt{r}} \sum_{s=0}^{r-1} e^{\frac{2i\pi sk}{r}} |u_s\rangle = \frac{1}{r} \sum_{s=0}^{r-1} e^{\frac{2i\pi sk}{r}} \sum_{j=0}^{r-1} e^{-\frac{2i\pi sj}{r}} |x^j \mod N\rangle 
= \frac{1}{r} \sum_{j=0}^{r-1} (\sum_{s=0}^{r-1} (e^{\frac{2i\pi(k-j)}{r}})^s) |x^j \mod N\rangle 
= \frac{1}{r} \sum_{j=0}^{r-1} r \delta_{jk} |x^j \mod N\rangle 
= |x^k \mod N\rangle$$

For k = 0 we obtain

$$\frac{1}{\sqrt{r}} \sum_{s=0}^{r-1} |u_s\rangle = |1\rangle$$

The easiest way is to think with the prime decomposition of the integers x and y. Let  $d = x \wedge y$  and  $m = x \vee y$ . Let  $p_0, p_1, \ldots, p_n$  be the prime numbers which appear in either prime decomposition. We can write

$$x = p_0^{\alpha_0} p_1^{\alpha_1} \dots p_n^{\alpha_n}$$
$$y = p_0^{\beta_0} p_1^{\beta_1} \dots p_n^{\beta_n}$$

where  $\alpha_i, \beta_i \in \mathbb{N}$ . Then it is clear that

$$d = p_0^{\gamma_0} p_1^{\gamma_1} \dots p_n^{\gamma_n}$$
$$m = p_0^{\delta_0} p_1^{\delta_1} \dots p_n^{\delta_n}$$

where  $\gamma_i = \min(\alpha_i, \beta_i)$  and  $\delta_i = \max(\alpha_i, \beta_i)$ . We have  $\alpha_i + \beta_i = \gamma_i + \delta_i$ . Then,

$$md = p_0^{\gamma_0} p_1^{\gamma_1} \dots p_n^{\gamma_n} p_0^{\delta_0} p_1^{\delta_1} \dots p_n^{\delta_n}$$

$$= p_0^{\gamma_0 + \delta_0} p_1^{\gamma_1 + \delta_1} \dots p_n^{\gamma_n + \delta_n}$$

$$= p_0^{\alpha_0 + \beta_0} p_1^{\alpha_1 + \beta_1} \dots p_n^{\alpha_n + \beta_n}$$

$$= xy$$

16.

Let  $x \ge 2$ .

$$\int_{x}^{x+1} \frac{1}{y^{2}} dy = \frac{1}{x} - \frac{1}{x+1}$$
$$= \frac{1}{x(x+1)}$$

since

$$x+1\leqslant \frac{3}{2}x \Leftrightarrow 2\leqslant x$$

$$\int_{x}^{x+1} \frac{1}{y^2} \, \mathrm{d}y = \frac{1}{x(x+1)} \geqslant \frac{2}{3x^2}$$

If we sum these inequalities

$$\sum_{q=2}^{+\infty} \frac{1}{q^2} \leqslant \frac{3}{2} \sum_{q=2}^{+\infty} \int_q^{q+1} \frac{1}{y^2} \, \mathrm{d}y = \frac{3}{2} \int_2^{+\infty} \frac{1}{y^2} \, \mathrm{d}y = \frac{3}{4}$$

and finally

$$\sum_{\substack{q \in \mathbb{N}^* \\ q \text{ is prime}}} \frac{1}{q^2} \leqslant \sum_{q=2}^{+\infty} \frac{1}{q^2} \leqslant \frac{3}{4}$$

17.

**17.1.** The assertion  $N=a^b\Rightarrow b\leqslant L$  is obviously wrong if N=a=1. Since we aim to prove an asymptotical result, we can assume that  $N\geqslant 2$ .

$$\begin{split} N &= a^b \Leftrightarrow \log N = b \log a \\ &\Leftrightarrow \frac{\log N}{\log a} = b & (N \geqslant 2 \Rightarrow a \geqslant 2 \Rightarrow \log a \geqslant 1 > 0) \\ &\Rightarrow b \leqslant \log N \\ &\Leftrightarrow b \leqslant \lfloor \log N \rfloor = L - 1 < L & (b \in \mathbb{N}) \end{split}$$

**17.2.** Let 
$$N = 2^l + a_{l-1}2^{l-1} + \dots + a_12 + a_0$$
 with  $l+1 \le L$  and  $a_i \in \{0,1\}$ .

$$N = 2^{l}(1 + a_{l-1}2^{-1} + \dots + a_{1}2^{-l+1} + a_{0}2^{-l})$$
  
=  $2^{l}(1 + f)$ 

with  $f \in [0, 1[$ .

$$\log N = l + \log(1 + a_{l-1}2^{-1} + \dots + a_12^{-l+1} + a_02^{-l})$$
  
=  $l + \log(1 + f)$ 

where log is  $\log_2$ . This shows that to compute an approximation to  $\log N$ , we just need an approximation of log in range [1, 2[ or any interval of the form [t, 2t[ for instance  $[\frac{3}{4}, \frac{1}{2}[$ . Besides,

$$\forall x \in ]-1,1], \quad \ln(1+x) = x - \frac{x^2}{2} + \frac{x^3}{3} + \dots + (-1)^{n+1} \frac{x^n}{n} + \dots$$
$$= \sum_{k=1}^{+\infty} (-1)^{k+1} \frac{x^k}{k}$$

Let's write it until order L-1:

$$\forall x \in ]-1, +\infty[, \quad \ln(1+x) = x - \frac{x^2}{2} + \frac{x^3}{3} + \dots + (-1)^{L+1} \frac{x^{L-1}}{L-1} + \sum_{k=L}^{+\infty} (-1)^{k+1} \frac{x^k}{k}$$

For  $x \in [0, \frac{1}{2}[$ , this is an alternating series and we can bound the rest by

$$\left| \sum_{k=L}^{+\infty} (-1)^{k+1} \frac{x^k}{k} \right| \leqslant \left| (-1)^L \frac{x^L}{L} \right|$$

$$\leqslant \frac{1}{2^L L}$$

For  $x \in [-\frac{1}{4}, 0]$ , we can use Lagrange formula to bound the rest by

$$\exists \xi \in [-\frac{1}{4}, 0[, \quad |\sum_{k=L}^{+\infty} (-1)^{k+1} \frac{x^k}{k}| = |\frac{\log^{(L)}(\xi)}{(L)!} x^L|$$

$$= \frac{(L-1)!}{(1+\xi)^L L!} |x^L|$$

$$= \frac{1}{(1+\xi)^L L} |x^L|$$

$$\leqslant \frac{1}{(1+\xi)^L L} \frac{1}{4^L}$$

$$\leqslant \frac{1}{(\frac{3}{4})^L L} \frac{1}{4^L}$$

$$= \frac{1}{3^L L}$$

This shows that we can use the Taylor series up to order L-1 to approximate  $\ln(x)$  with precision  $2^{-L}$  on the range  $\left[\frac{3}{4},\frac{1}{2}\right]$ . This is to simplify the complexity analysis. In actual implementation though better and faster approximation are used: See the book by Cheney [4] for mathematical fundations of the approximation of functions by polynomials including the Remez algorithm. See also this insightful post [3] which discusses tradeoffs between accuracy and speed in approximating this log function, taking into account error induced by floating-point representation of real numbers. Here [5] can be found an actual implementation of the C standard library.

In addition to the Taylor error, there is an error occurring when computing the polynomial using floating-point arithmetic. If we store the significand of the floating-point variables in binary on L+1 bits, and use

O(L) bits to do arithmetic operations, each operation will incurr a relative error of at most  $\epsilon = 2^{-L-1}$ , i.e.

$$x \oplus y = (x+y)(1+\xi)$$
$$x \ominus y = (x-y)(1+\xi)$$
$$x \otimes y = (x \times y)(1+\xi)$$
$$x \oslash y = (x \div y)(1+\xi)$$

where  $|\xi| \leq \epsilon$  and the values on the left are the value computed exactly and then rounded on L+1 digits. For the details on floating point arithmetic see [6]. The previous polynomial can be rewritten as:

$$P(x) = \sum_{k=1}^{n} (-1)^{k+1} \frac{1}{k} x^k$$
  
=  $x(1 + x(-\frac{1}{2} + x(\frac{1}{3} + \dots + ((-1)^{n-1} \frac{1}{n-1} + (-1)^n \frac{1}{n} x) \dots)))$ 

This shows that the evaluation costs n fused multiply-add operations. If one rounding error occurs for each of the multiply-add, we have the following bound on the error due to floating-point arithmetic (see [7] for a detailed analysis)):

$$|\bar{P}(x) - \tilde{P}(x)| = |\sum_{j=1}^{n-1} (\xi_j \sum_{i=j}^n (-1)^{i+1} \frac{1}{i} x^i)|$$

$$= |\sum_{i=1}^{n-1} (\sum_{j=1}^i \xi_j) (-1)^{i+1} \frac{1}{i} x^i + (\sum_{j=1}^{n-1} \xi_j) (-1)^n \frac{1}{n} x^n|$$

$$\leq \sum_{i=1}^{n-1} (\sum_{j=1}^i \epsilon) \frac{1}{i} |x^i| + (\sum_{j=1}^{n-1} \epsilon) \frac{1}{n} |x^n|$$

$$= \epsilon \sum_{i=1}^{n-1} |x^i| + \epsilon \frac{n-1}{n} |x^n|$$

$$\leq \epsilon \sum_{i=1}^n \frac{1}{2^i}$$

$$= \epsilon (1 - (\frac{1}{2})^n)$$

$$\leq \epsilon$$

and we add the error due to just storing the coefficients of the polynomial on L bits: for instance  $\frac{1}{3} = 0.010101...$  is rounded when storing in binary. If  $\tilde{a}_i$  is the rounded value of  $a_i = (-1)^i \frac{1}{i}$ , the error will be:

$$|P(x) - \tilde{P}(x)| \leqslant \sum_{i=1}^{n} |a_i - \tilde{a}_i| |x^i|$$

$$\leqslant \sum_{i=1}^{n} \epsilon |a_i| |x^i|$$

$$= \epsilon \sum_{i=1}^{n} \frac{1}{i} |x^i|$$

$$\leqslant \epsilon \sum_{i=1}^{n} |x^i|$$

$$\leqslant \epsilon$$

Taking into consideration the three types of error, we see that the taylor series of order L is a approximation to  $\ln(x)$  on range  $\left[\frac{3}{4}, \frac{1}{2}\right[$  with precision  $2^{-L}$  since :

$$\frac{1}{2^{L+1}(L+1)} + 2\epsilon \leqslant 2^{-L}$$

$$\Leftrightarrow \frac{2^{-L-1}}{L} + 2^{-L} \leqslant 2^{-L}$$

$$\Leftrightarrow 2^{-L-1}(\frac{1}{L} + 1) \leqslant 2^{-L}$$

$$\Leftrightarrow L \geqslant 1$$

This analysis shows that the procedure Log2 computes an approximation of  $\log(N)$  to precision  $2^{-L}$ . Binary addition-substraction costs  $\Theta(L)$  operations, grade-school multiplication-division costs  $\Theta(L^2)$ . Multiplication complexity can be improved to:

- $O(L^{\log_2(3)})$  using Karatsuba algorithm [2].
- $-O(L\log(L)\log\log(L))$  using Schönhage-Strassen algorithm [2].
- $O(L \log L \log^* L \text{ using Furer algorithm [8]}.$

Faster division  $x \div y$  consists in computing  $\frac{1}{y}$  in  $O(\log(L))$  multiplications, then doing  $x \div y = x \times \frac{1}{y}$  (cf. [9]). In the end computing  $\log_2 N$  has an  $O(L^3)$  time complexity. If we are given an approximating polynomial and are assured it gives the desired precision for any input size considered, the complexity is  $O(L^2)$ . The complexity of finding  $|\log_2(N)|$  given the binary representation of N is O(L).

```
 \begin{aligned} & Log2(N,L) \\ & \# L \geqslant l+1 \text{ where } l = \lfloor \log_2(N) \rfloor, \text{ i.e. } 2^l \leqslant N < 2^{l+1}. \\ & \textbf{for } j = 1 \textbf{ to } L \\ & A[j] = (-1)^{j+1} \frac{1}{j} \\ & m = \lfloor \log_2(N) \rfloor & \# N = 2^m (1+f) \\ & f = \frac{N}{2^m} - 1 & \# \text{ no rounding error in } f. \\ & \textbf{if } f \geqslant \frac{1}{2} & \# \text{ map range } [1.5, 2[ \text{ to } [0.75, 1[.\\ & f = \frac{1/2 - f}{2} \\ & m = m + 1 \\ & q = 0 \\ & \textbf{for } j = L \textbf{ downto } 0 \\ & q = q \times f + A[j] \\ & q = q \div \ln(2) \\ & \textbf{return } q + m \end{aligned}
```

18.

$$x^{2} = 16$$
  
 $x^{3} = 64 = -27$   
 $x^{4} = 108 = 17$   
 $x^{5} = 68 = -23$   
 $x^{6} = 92 = 1$ 

shows that the order of x is 6.

19.

1 is not composite and all the odd integers less than 15 are prime except  $9 = 3^2$ .

Let  $l \in [0, N-1]$ .

$$\sum_{x=0}^{N-1} e^{-\frac{2i\pi lx}{N}} f(x) = \sum_{k=0}^{\frac{N}{r}-1} \sum_{x=0}^{r-1} e^{-\frac{2i\pi l(kr+x)}{N}} f(kr+x)$$

$$= \sum_{k=0}^{\frac{N}{r}-1} \sum_{x=0}^{r-1} e^{-\frac{2i\pi lkr}{N}} e^{-\frac{2i\pi lx}{N}} f(x)$$

$$= \sum_{x=0}^{r-1} (\sum_{k=0}^{\frac{N}{r}-1} e^{-\frac{2i\pi lkr}{N}}) e^{-\frac{2i\pi lx}{N}} f(x)$$

$$= \sum_{x=0}^{r-1} (\sum_{k=0}^{\frac{N}{r}-1} e^{-\frac{2i\pi lkr}{N}}) e^{-\frac{2i\pi lx}{N}} f(x)$$

we have

$$\sum_{k=0}^{\frac{N}{r}-1} e^{-\frac{2i\pi lkr}{N}} = \begin{cases} \frac{N}{r} & \text{if } \frac{lr}{N} \in \mathbb{N}, \\ \frac{1 - (e^{-\frac{2i\pi lr}{N}})^{\frac{N}{r}}}{1 - e^{-\frac{2i\pi lr}{N}}} = 0 & \text{otherwise.} \end{cases}$$

so if  $l = l' \frac{N}{r}$ , with  $l' \in [0, r - 1]$ ,

$$\frac{1}{\sqrt{N}} \sum_{x=0}^{N-1} e^{-\frac{2i\pi lx}{N}} f(x) = \sqrt{\frac{N}{r}} \frac{1}{\sqrt{r}} \sum_{x=0}^{r-1} e^{-\frac{2i\pi l'x}{r}} f(x)$$

otherwise

$$\frac{1}{\sqrt{N}} \sum_{x=0}^{N-1} e^{-\frac{2i\pi lx}{N}} f(x) = 0$$

**22**.

First we observe that because of the periodicity, we have

$$f(x_1, x_2) = f(x_1 - 1, x_2 + s)$$

$$= \dots$$

$$= f(1, x_2 + (x_1 - 1)s)$$

$$= f(0, x_2 + x_1s)$$

Second, for a fixed  $x_1$ , the application

$$\varphi_{x_1}: \llbracket 0, r-1 \rrbracket \to \llbracket 0, r-1 \rrbracket$$
 
$$x_2 \mapsto x_2 + x_1 s \mod r$$

is a permutation of [0, r-1], i.e. for each  $j \in [0, r-1]$ , there is exactly one  $x_2$  such that  $x_2 + x_1s = j \mod r$ . Finally, if  $x_2 + x_1s = j \mod r$ ,

$$e^{-\frac{2i\pi(l_1x_1+l_2x_2)}{r}} = e^{-\frac{2i\pi(l_1x_1+l_2(j-x_1s+kr))}{r}}$$
$$= e^{-\frac{2i\pi((l_1-l_2s)x_1+l_2j)}{r}}$$



FIGURE 4. Periodicity of f with r = 15 and s = 4.

Let's now rewrite the sum. Let  $l_1, l_2 \in [0, r-1]$ :

$$|\hat{f}(l_1, l_2)\rangle = \sum_{x_1, x_2=0}^{r-1} e^{-\frac{2i\pi(l_1x_1 + l_2x_2)}{r}} |f(x_1, x_2)\rangle$$

$$= \sum_{x_1=0}^{r-1} \sum_{x_2=0}^{r-1} e^{-\frac{2i\pi(l_1x_1 + l_2x_2)}{r}} |f(x_1, x_2)\rangle$$

$$= \sum_{x_1=0}^{r-1} e^{-\frac{2i\pi l_1x_1}{r}} \sum_{x_2=0}^{r-1} e^{-\frac{2i\pi l_2x_2}{r}} |f(0, x_2 + x_1s)\rangle$$

$$= \sum_{x_1=0}^{r-1} e^{-\frac{2i\pi l_1x_1}{r}} \sum_{j=0}^{r-1} e^{-\frac{2i\pi l_2(-x_1s+j)}{r}} |f(0, j)\rangle$$

$$= \sum_{x_1=0}^{r-1} e^{-\frac{2i\pi(l_1 - l_2s)x_1}{r}} \sum_{j=0}^{r-1} e^{-\frac{2i\pi l_2j}{r}} |f(0, j)\rangle$$

$$= \sum_{x_1=0}^{r-1} (e^{-\frac{2i\pi(l_1 - l_2s)}{r}})^{x_1} \sum_{j=0}^{r-1} e^{-\frac{2i\pi l_2j}{r}} |f(0, j)\rangle$$

Using the usual argument, the first factor which is a geometric sum is 0 unless  $l_1 - l_2 s = kr$  with  $k \in \mathbb{Z}$  and in that case

$$\sum_{x_1, x_2=0}^{r-1} e^{-\frac{2i\pi(l_1x_1+l_2x_2)}{r}} |f(x_1, x_2)\rangle = r \sum_{j=0}^{r-1} e^{-\frac{2i\pi l_2j}{r}} |f(0, j)\rangle$$

23.

Let  $x_1, x_2 \in [0, r-1]$ .

$$\sum_{l_1, l_2 = 0}^{r-1} e^{\frac{2i\pi(x_1 l_1 + x_2 l_2)}{r}} |\hat{f}(l_1, l_2)\rangle = \sum_{l_2 = 0}^{r-1} \sum_{l_1 = 0}^{r-1} e^{\frac{2i\pi(x_1 l_1 + x_2 l_2)}{r}} |\hat{f}(l_1, l_2)\rangle$$

for a given value of  $l_2$ , there is exactly one  $l_1 \in [0, r-1]$  such that  $l_1 = l_2 s \mod r$ , so there is exactly one term in each inner sum which is non zero and

$$\begin{split} \sum_{l_2=0}^{r-1} \sum_{l_1=0}^{r-1} e^{\frac{2i\pi(x_1l_1+x_2l_2)}{r}} |\hat{f}(l_1,l_2)\rangle &= r \sum_{l_2=0}^{r-1} e^{\frac{2i\pi(x_1l_2s+x_2l_2)}{r}} \sum_{j=0}^{r-1} e^{-\frac{2i\pi l_2j}{r}} |f(0,j)\rangle \\ &= r \sum_{j=0}^{r-1} (\sum_{l_2=0}^{r-1} e^{\frac{2i\pi(x_1l_2s+x_2l_2-jl_2)}{r}}) |f(0,j)\rangle \\ &= r \sum_{j=0}^{r-1} (\sum_{l_2=0}^{r-1} (e^{\frac{2i\pi(x_1s+x_2-j)}{r}})^{l_2}) |f(0,j)\rangle \end{split}$$

Again a geometric sum

$$\sum_{l_2=0}^{r-1} \left(e^{\frac{2i\pi(x_1s+x_2-j)}{r}}\right)^{l_2} = \begin{cases} r & \text{if } j = x_2+x_1s \mod r, \\ 0 & \text{otherwise.} \end{cases}$$

So finally

$$\begin{split} \sum_{l_2=0}^{r-1} \sum_{l_1=0}^{r-1} e^{\frac{2i\pi(x_1l_1+x_2l_2)}{r}} |\hat{f}(l_1,l_2)\rangle &= r^2 |f(0,x_2+x_1s+kr)\rangle \\ &= r^2 |f(0,x_2+x_1s)\rangle \\ &= r^2 |f(x_1,x_2)\rangle \end{split}$$

26.

Any finite abelian group is a direct sum of cyclic groups of prime power order (cf. [10] or these shorter notes [11]).

$$G \cong \mathbb{Z}/p_1^{\beta_1}\mathbb{Z} \times \mathbb{Z}/p_2^{\beta_2}\mathbb{Z} \times \cdots \times \mathbb{Z}/p_n^{\beta_n}\mathbb{Z}$$

Such a decomposition of the group G does not necessarily imply a similar decomposition of subgroup K as a product of subgroup. Let's take the example of  $G = \mathbb{Z}/2\mathbb{Z} \times \mathbb{Z}/8\mathbb{Z}$  and K = <(1,2)>: not only K is not a product but there exists no isomorphism of G which maps K into the desired form (cf. [12]). We have to make the additional assumption that K is decomposed as a product of subgroups of the direct factors of G:

$$K = p_1^{\alpha_1} \mathbb{Z}/p_1^{\beta_1} \mathbb{Z} \times p_2^{\alpha_2} \mathbb{Z}/p_2^{\beta_2} \mathbb{Z} \times \dots \times p_n^{\alpha_n} \mathbb{Z}/p_n^{\beta_n} \mathbb{Z}$$

where  $\alpha_i \leq \beta_i$ . Note that

$$|K| = p_1^{\beta_1 - \alpha_1} \times p_2^{\beta_2 - \alpha_2} \times \dots \times p_n^{\beta_n - \alpha_n}$$

and that we then have the following decomposition for the quotient of group G by it normal subgroup K:

$$G/K = \mathbb{Z}/p_1^{\alpha 1} \mathbb{Z} \times \mathbb{Z}/p_2^{\alpha 2} \mathbb{Z} \times \dots \times \mathbb{Z}/p_n^{\alpha n} \mathbb{Z}$$

Let  $l = (l_1, l_2, ..., l_n) \in \mathbb{Z}/p_1^{\beta_1}\mathbb{Z} \times \mathbb{Z}/p_2^{\beta_2}\mathbb{Z} \times \cdots \times \mathbb{Z}/p_n^{\beta_n}\mathbb{Z}$ . We define  $|\hat{f}(l)\rangle$  by:

$$\begin{split} \sqrt{|G||K|} \, |\hat{f}(l)\rangle &= \sum_{g \in G} e^{-2i\pi (l_1 \frac{g_1^{\beta}}{p_1^{\beta}} + l_2 \frac{g_2^{\beta}}{p_2^{\beta}} + \dots + l_n \frac{g_n}{p_n^{\beta_n}})} \, |f(g_1, g_2, \dots, g_n)\rangle \\ &= \sum_{k \in K} \sum_{x \in G/K} e^{-2i\pi (l_1 \frac{k_1 + x_1}{p_1^{\beta_1}} + l_2 \frac{k_2 + x_2}{p_2^{\beta_2}} + \dots + l_n \frac{k_n + x_n}{p_n^{\beta_n}})} \, |f(k_1 + x_1, k_2 + x_2, \dots, k_n + x_n)\rangle \\ &= \sum_{k \in K} e^{-2i\pi (l_1 \frac{k_1}{p_1^{\beta_1}} + l_2 \frac{k_2}{p_2^{\beta_2}} + \dots + l_n \frac{k_n}{p_n^{\beta_n}})} \sum_{x \in G/K} e^{-2i\pi (l_1 \frac{x_1}{p_1^{\beta_1}} + l_2 \frac{x_2}{p_2^{\beta_2}} + \dots + l_n \frac{k_n}{p_n^{\beta_n}})} \sum_{x \in G/K} e^{-2i\pi (l_1 \frac{x_1}{p_1^{\beta_1}} + l_2 \frac{x_2}{p_2^{\beta_2}} + \dots + l_n \frac{x_n}{p_n^{\beta_n}})} \, |f(k_1 + x_1, k_2 + x_2, \dots, k_n + x_n)\rangle \\ &= \sum_{k \in K} e^{-2i\pi l_1 \frac{k_1}{p_1^{\beta_1}} + l_2 \frac{k_2}{p_2^{\beta_2}}} \times \dots \times e^{-2i\pi l_n \frac{k_n}{p_n^{\beta_n}}} \sum_{x \in G/K} e^{-2i\pi (l_1 \frac{x_1}{p_1^{\beta_1}} + l_2 \frac{x_2}{p_2^{\beta_2}} + \dots + l_n \frac{x_n}{p_n^{\beta_n}})} \, |f(x_1, x_2, \dots, x_n)\rangle \\ &= (\sum_{k_1 = 0}^{p_1^{\beta_1 - \alpha_1} - 1} e^{-2i\pi l_1 \frac{k_1 p_1^{\alpha_1}}{p_1^{\beta_1}}}) (\sum_{k_2 = 0}^{p_2^{\beta_2 - \alpha_2} - 1} e^{-2i\pi l_2 \frac{k_2 p_2^{\alpha_2}}{p_2^{\beta_2}}}) \times \dots \times (\sum_{k_n = 0}^{p_n^{\beta_n - \alpha_n} - 1} e^{-2i\pi l_n \frac{k_n p_n^{\alpha_n}}{p_n^{\beta_n}}}) \\ &= (\sum_{k_1 = 0}^{p_1^{\beta_1 - \alpha_1} - 1} (e^{-2i\pi l_1 \frac{p_1^{\alpha_1}}{p_1^{\beta_1}}})^{k_1}) (\sum_{k_2 = 0}^{p_2^{\beta_2 - \alpha_2} - 1} (e^{-2i\pi l_2 \frac{p_2^{\alpha_2}}{p_2^{\beta_2}}})^{k_2}) \times \dots \times (\sum_{k_n = 0}^{p_n^{\beta_n - \alpha_n} - 1} (e^{-2i\pi l_n \frac{p_n^{\alpha_n}}{p_n^{\beta_n}}})^{k_n}) \\ &\times \sum_{x \in G/K} e^{-2i\pi (l_1 \frac{x_1}{p_1^{\beta_1}} + l_2 \frac{x_2}{p_2^{\beta_2}} + \dots + l_n \frac{x_n}{p_n^{\beta_n}})} \, |f(x_1, x_2, \dots, x_n)\rangle \end{aligned}$$

where we have used the fact that f is constant on the cosets of K. The n first factors are geometric sums, everyone of them is non zero if and only if

$$(l_1, l_2, \dots, l_n) = (l'_1 p_1^{\beta_1 - \alpha_1}, l'_2 p_2^{\beta_2 - \alpha_2}, \dots, l'_n p_n^{\beta_n - \alpha_n})$$

with  $(l'_1, l'_2, \dots, l'_n) \in G/K$ . In that case the last sum becomes:

$$\sum_{x \in G/K} e^{-2i\pi (l_1' \frac{x_1}{p_1^{\alpha_1}} + l_2' \frac{x_2}{p_2^{\alpha_2}} + \dots + l_n' \frac{x_n}{p_n^{\alpha_n}})} |f(x_1, x_2, \dots, x_n)\rangle$$

Since the values of f on 2 different cosets are distincts, the family of vectors  $|\hat{f}\rangle$  is an orthonormal basis of the  $\frac{|G|}{|K|}$ -dimensional subspace spanned by the vectors  $|f(x_1, x_2, \dots, x_n)\rangle$ . Inverting these equalities allows us to write

$$\forall (x_1, x_2, \dots, x_n) \in G, \quad |f(x_1, x_2, \dots, x_n)\rangle = \sqrt{\frac{|K|}{|G|}} \sum_{l' \in G/K} e^{2i\pi(l'_1 \frac{x_1}{p_1^{\alpha_1}} + l'_2 \frac{x_2}{p_2^{\alpha_2}} + \dots + l'_n \frac{x_n}{p_n^{\alpha_n}})} |\hat{f}(l'_1 p_1^{\beta_1 - \alpha_1}, l'_2 p_2^{\beta_2 - \alpha_2}, \dots, l'_n p_n^{\beta_n - \alpha_n})\rangle$$

Applying the phase estimation algorithm will thus allow to determine

$$(\underbrace{\frac{\widetilde{l_1'}}{p_1^{\alpha_1}}, \underbrace{\widetilde{l_2'}}_{p_2^{\alpha_2}}, \dots, \underbrace{\frac{\widetilde{l_n'}}{p_n^{\alpha_n}}})$$

and the continued fraction algorithm will provide  $(p_1^{\alpha_1}, p_2^{\alpha_2}, \dots, p_n^{\alpha_n})$  which is all we need to characterize the hidden subgroup K.

### 28.

We start with the initial state

$$\underbrace{|0\rangle |0\rangle \dots |0\rangle}_{t=t_1+t_2+\dots+t_n \text{ bits}} \underbrace{|0\rangle}_{t=t_1+t_2+\dots+t_n \text{ bits}}$$

We create the superposition

$$\frac{1}{\sqrt{2^t}} \sum_{x_1=0}^{2^{t_1}-1} \sum_{x_2=0}^{2^{t_2}-1} \cdots \sum_{x_n=0}^{2^{t_n}-1} |x_1\rangle |x_2\rangle \dots |x_n\rangle |0\rangle$$

f is a function from G to  $[0, 2^m - 1]$  which is constant on the cosets of K, such that the values on 2 different cosets are distincts:

$$f: \mathbb{Z}/p_1^{\beta_1}\mathbb{Z} \times \mathbb{Z}/p_2^{\beta_2}\mathbb{Z} \times \dots \times \mathbb{Z}/p_n^{\beta_n}\mathbb{Z} \to [0, 2^{m-1}]$$
$$(g_1, g_2, \dots, g_n) \mapsto f(g_1, g_2, \dots, g_n)$$

We definine  $\tilde{f}$  to be an extension of the function f:

$$\tilde{f}: [0, 2^{t_1} - 1] \times [0, 2^{t_2} - 1] \times \dots [0, 2^{t_n} - 1] \to [0, 2^{m-1}]$$

$$(x_1, x_2, \dots, x_n) \mapsto f(x_1 \mod p_1^{\beta_1}, x_2 \mod p_2^{\beta_2}, \dots, x_n \mod p_n^{\beta_n})$$

We apply the operator

$$U: |x_1\rangle |x_2\rangle \dots |x_n\rangle |y\rangle \mapsto |x_1\rangle |x_2\rangle \dots |x_n\rangle |y \oplus \tilde{f}(x_1, x_2, \dots, x_n)\rangle$$

to get the state

$$\frac{1}{\sqrt{2^t}} \sum_{x_1=0}^{2^{t_1}-1} \sum_{x_2=0}^{2^{t_2}-1} \cdots \sum_{x_n=0}^{2^{t_n}-1} |x_1\rangle |x_2\rangle \dots |x_n\rangle |\tilde{f}(x_1, x_2, \dots, x_n)\rangle$$

We can express the  $|f\rangle$  in the  $|\hat{f}\rangle$  basis:

$$\begin{split} &\frac{1}{\sqrt{2^{t}}}\sqrt{\frac{|K|}{|G|}}\sum_{x_{1}=0}^{2^{t_{1}}-1}\sum_{x_{2}=0}^{2^{t_{2}}-1}\cdots\sum_{x_{n}=0}^{2^{t_{n}}-1}|x_{1}\rangle\,|x_{2}\rangle\,\ldots\,|x_{n}\rangle\,\sum_{l'\in G/K}e^{2i\pi(l'_{1}\frac{x_{1}}{p_{1}^{\alpha_{1}}}+l'_{2}\frac{x_{2}}{p_{2}^{\alpha_{2}}}+\cdots+l'_{n}\frac{x_{n}}{p_{n}^{\alpha_{n}}})}\,|\hat{f}(l'_{1}p_{1}^{\beta_{1}-\alpha_{1}},l'_{2}p_{2}^{\beta_{2}-\alpha_{2}},\ldots,l'_{n}p_{n}^{\beta_{n}-\alpha_{n}})\rangle\\ &=\sqrt{\frac{|K|}{|G|}}\sum_{l'\in G/K}(\frac{1}{\sqrt{2^{t}}}\sum_{x_{1}=0}^{2^{t_{1}}-1}\sum_{x_{2}=0}^{2^{t_{2}}-1}\cdots\sum_{x_{n}=0}^{2^{t_{n}}-1}e^{2i\pi(l'_{1}\frac{x_{1}}{p_{1}^{\alpha_{1}}}+l'_{2}\frac{x_{2}}{p_{2}^{\alpha_{2}}}+\cdots+l'_{n}\frac{x_{n}}{p_{n}^{\alpha_{n}}})}\,|x_{1}\rangle\,|x_{2}\rangle\,\ldots\,|x_{n}\rangle)\,|\hat{f}(l'_{1}p_{1}^{\beta_{1}-\alpha_{1}},l'_{2}p_{2}^{\beta_{2}-\alpha_{2}},\ldots,l'_{n}p_{n}^{\beta_{n}-\alpha_{n}})\rangle\\ &=\sqrt{\frac{|K|}{|G|}}\sum_{l'\in G/K}(\frac{1}{\sqrt{2^{t_{1}}}}\sum_{x_{1}=0}^{2^{t_{1}}-1}e^{2i\pi l'_{1}\frac{x_{1}}{p_{1}^{\alpha_{1}}}}\,|x_{1}\rangle)(\frac{1}{\sqrt{2^{t_{2}}}}\sum_{x_{2}=0}^{2^{t_{2}}-1}e^{2i\pi l'_{2}\frac{x_{2}}{p_{2}^{\alpha_{2}}}}\,|x_{2}\rangle)\ldots(\frac{1}{\sqrt{2^{t_{n}}}}\sum_{x_{1}=0}^{2^{t_{n}}-1}e^{2i\pi l'_{n}\frac{x_{n}}{p_{n}^{\alpha_{n}}}}\,|x_{n}\rangle)\,|\hat{f}(l'_{1}p_{1}^{\beta_{1}-\alpha_{1}},l'_{2}p_{2}^{\beta_{2}-\alpha_{2}},\ldots,l'_{n}p_{n}^{\beta_{2}-\alpha_{2}},\ldots,l'_{n}p_{n}^{\beta_{2}-\alpha_{2}},\ldots,l'_{n}p_{n}^{\beta_{2}-\alpha_{2}},\ldots,l'_{n}p_{n}^{\beta_{2}-\alpha_{2}},\ldots,l'_{n}p_{n}^{\beta_{2}-\alpha_{2}},\ldots,l'_{n}p_{n}^{\beta_{2}-\alpha_{2}},\ldots,l'_{n}p_{n}^{\beta_{2}-\alpha_{2}},\ldots,l'_{n}p_{n}^{\beta_{2}-\alpha_{2}},\ldots,l'_{n}p_{n}^{\beta_{2}-\alpha_{2}},\ldots,l'_{n}p_{n}^{\beta_{2}-\alpha_{2}},\ldots,l'_{n}p_{n}^{\beta_{2}-\alpha_{2}},\ldots,l'_{n}p_{n}^{\beta_{2}-\alpha_{2}},\ldots,l'_{n}p_{n}^{\beta_{2}-\alpha_{2}},\ldots,l'_{n}p_{n}^{\beta_{2}-\alpha_{2}},\ldots,l'_{n}p_{n}^{\beta_{2}-\alpha_{2}},\ldots,l'_{n}p_{n}^{\beta_{2}-\alpha_{2}},\ldots,l'_{n}p_{n}^{\beta_{2}-\alpha_{2}},\ldots,l'_{n}p_{n}^{\beta_{2}-\alpha_{2}},\ldots,l'_{n}p_{n}^{\beta_{2}-\alpha_{2}},\ldots,l'_{n}p_{n}^{\beta_{2}-\alpha_{2}},\ldots,l'_{n}p_{n}^{\beta_{2}-\alpha_{2}},\ldots,l'_{n}p_{n}^{\beta_{2}-\alpha_{2}},\ldots,l'_{n}p_{n}^{\beta_{2}-\alpha_{2}},\ldots,l'_{n}p_{n}^{\beta_{2}-\alpha_{2}},\ldots,l'_{n}p_{n}^{\beta_{2}-\alpha_{2}},\ldots,l'_{n}p_{n}^{\beta_{2}-\alpha_{2}},\ldots,l'_{n}p_{n}^{\beta_{2}-\alpha_{2}},\ldots,l'_{n}p_{n}^{\beta_{2}-\alpha_{2}},\ldots,l'_{n}p_{n}^{\beta_{2}-\alpha_{2}},\ldots,l'_{n}p_{n}^{\beta_{2}-\alpha_{2}},\ldots,l'_{n}p_{n}^{\beta_{2}-\alpha_{2}},\ldots,l'_{n}p_{n}^{\beta_{2}-\alpha_{2}},\ldots,l'_{n}p_{n}^{\beta_{2}-\alpha_{2}},\ldots,l'_{n}p_{n}^{\beta_{2}-\alpha_{2}},\ldots,l'_{n}p_{n}^{\beta_{2}-\alpha_{2}},\ldots,l'_{n}p_{n}^{\beta_{2}-\alpha_{2}},\ldots,l'_{n}p_$$

We apply inverse fourier transform on the first n registers:

$$\sqrt{\frac{|K|}{|G|}} \sum_{l' \in G/K} |\widetilde{\frac{l'_1}{p_1^{\alpha_1}}}\rangle |\widetilde{\frac{l'_2}{p_2^{\alpha_2}}}\rangle \dots |\widetilde{\frac{l'_n}{p_n^{\alpha_n}}}\rangle |\widehat{f}(l'_1 p_1^{\beta_1 - \alpha_1}, l'_2 p_2^{\beta_2 - \alpha_2}, \dots, l'_n p_n^{\beta_n - \alpha_n})\rangle$$

We measure the first n registers to get

$$(\widetilde{\frac{l_1'}{p_1^{\alpha_1}}},\widetilde{\frac{l_2'}{p_2^{\alpha_2}}},\ldots,\widetilde{\frac{l_n'}{p_n^{\alpha_n}}})$$

The continued fraction algorithm provides

$$(p_1^{\alpha_1}, p_2^{\alpha_2}, \dots, p_n^{\alpha_n})$$

In register i, if we use  $t_i = 2\lceil \log p_i^{\beta i} \rceil + 1 + \lceil \log(2 + \frac{1}{2\epsilon}) \rceil$ , with probability at least  $1 - \epsilon$ , we will have some an approximation of  $\frac{l_i'}{p_i^{\alpha_i}}$  accurate to  $2\lceil \log p_i^{\beta i} \rceil + 1$  bits, for some  $l_i' \in \mathbb{Z}/p_i^{\alpha_i}\mathbb{Z}$ . The continued fraction algorithm will then succeed in providing  $p_i^{\alpha_i}$  unless  $l_i' \in p_i\mathbb{Z}$ , which happens with probability  $\frac{p_i^{\alpha_i-1}}{p_i^{\alpha_i}} = \frac{1}{p_i}$ . A lower bound on the probability that the algorithm succeeds in finding K is thus:

$$(1-\epsilon)^n(1-\frac{1}{p_1})(1-\frac{1}{p_2})\dots(1-\frac{1}{p_n})$$

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