

# Diffusion process as a computational engine: integer factorisation algorithm

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- 1 The diffusion as a (theoretical) computational engine
- 2 Random walks on regular graphs. Diffusion step.
- 3 Integer factorization

# Heat diffusion

Assume that a heat source, such as a flame or a welding torch, is applied to the center of a circular disc of uniform thickness and material composition.

Then two observers who are measuring temperatures at different points on the perimeter will detect a change of temperature at their points of contact at the same rate.

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Certain aspects of diffusion are simultaneously, not sequentially, observable!

# Diffusion of light

Within a circular ring, imagine a beam of light  $\mathcal{B}$  (or some type of focused energy) emanating from a source at a perimeter point  $\mathcal{P}_0$ . Upon contact with another perimeter point  $\mathcal{P}_1$  on the ring, the beam  $\mathcal{B}$  splits into  $M$  sub-beams of equal magnitude in a prescribed set of directions toward perimeter points  $\mathcal{P}_{2,1}, \dots, \mathcal{P}_{2,M}$ .

Let each sub-beam upon contact with some  $\mathcal{P}_{2,k}$  split in manner as similar to the reflection of  $\mathcal{B}$  at  $\mathcal{P}_1$ , and so on.

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Let each sub-beam upon contact with some  $\mathcal{P}_{2,k}$  split in manner as similar to the reflection of  $\mathcal{B}$  at  $\mathcal{P}_1$ , and so on.

Then after  $n$  such splittings, what portion of the original amount energy has returned to  $\mathcal{P}_0$ ?

## Diffusion step - intuitive definition

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In the above example: Each contact involves the splitting of a single beam into  $M$  sub-beams, then after  $n$  diffusion steps, one will have  $M^n$  paths of light traversing the ring.

Imagine a ring were 1 kilometer in diameter, and if the beam were to travel at the speed of light, then after 0.01 seconds one would expect to have more than  $M^{3000}$  sub-beams crossing various chords of the ring since in almost all circumstances more than 3000 diffusion steps would have taken place.



## A natural question

Q: What is the most natural mathematical setup for describing/measuring such diffusion?

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A: A Regular (weighted) graph

# Regular weighted graph

$X$  is undirected connected graph with:

- Finite set of vertices  $V$
- Set of edges  $E =$  collection of two element subsets of  $V$
- Real valued weight function  $w :: V \times V \rightarrow \mathbb{R}_{\geq 0}$  which is symmetric and  $w(x, y) > 0$  iff  $\{x, y \in E\}$ .

The degree of a vertex  $x \in V$  is

$$d(x) = \sum_{y \in V} w(x, y).$$

A weighted graph  $X$  is **regular** of degree  $d$  if  $d(x) = d$  for all vertices  $x \in V$ .

## Important example: Cayley graph

$G$  a finite abelian group

$S \subseteq G$  a fixed symmetric subset (sometimes one assumes it generates  $G$ ).

The symmetry condition means that  $s \in S$  iff  $-s \in S$  (or  $s \in S$  iff  $s^{-1} \in S$ ).

For any  $\alpha: S \rightarrow \mathbb{R}^{>0}$  such that  $\alpha(s) = \alpha(-s)$ , one can construct a weighted Cayley graph  $X = \text{Cay}(G, S, \alpha)$  of  $G$  with respect to  $S$  and  $\alpha$  as follows:

- The vertices of  $X$  are the elements of  $G$
- Two vertices  $x$  and  $y$  are connected with an edge if and only if  $x - y \in S$  (additive notation)
- The weight  $w(x, y)$  of the edge  $(x, y)$  is  $w(x, y) = \alpha(x - y)$

Then  $X = \text{Cay}(G, S, \alpha)$  is a regular weighted graph of degree  $d = \sum_{s \in S} \alpha(s)$ .

# Random walks

$X$  a regular weighted graph of degree  $d$

$A$  adjacency matrix (non-negative, symmetric!)

A half-lazy random walk on  $X$  is a Markov chain with state space  $(V, \mathcal{P}(V))$  with arbitrary initial probability distribution  $p_0: V \rightarrow \mathbb{R}$ , and the transition probability matrix

$$W := \frac{1}{2} \left( I + \frac{1}{d} A \right).$$

After  $n$  steps

$$p_n(x) = W^n p_0(x).$$

## Random walks - convergence

As expected, probability distribution  $p_n$  tends to the uniform distribution as  $n \rightarrow \infty$ .

The rate of convergence is:

$$\sup_{x \in V} \left| p_n(x) - \frac{1}{|V|} \right| \leq \lambda_1^n,$$

$\lambda_1$  is the largest eigenvalue of  $W$  less than 1.

Note that eigenvalues of  $W$  are nonnegative with the largest eigenvalue equal to 1 with multiplicity 1.

(Follows easily from the fact that  $A$  is symmetric with eigenvalues in  $[-d, d]$ .)

## "Diffusion computability"

A diffusion process in  $X = \text{Cay}(G, S, \alpha)$  may be regarded as an analog computation on  $X$

### Definition

A real-valued function  $h$  on  $G$  is said to be computable by a diffusion process in  $X$  with initial condition  $p_0: G \rightarrow \mathbb{R}$  if the following holds. Let  $\{p_m = W^m p_0\}_{m=0}^{\infty}$  be the sequence of distributions in  $X$  with initial probability distribution  $p_0$ . Then for any given  $\varepsilon > 0$  there exists a positive integer  $n = n(\varepsilon)$  such that for all  $m > n(\varepsilon)$  and all  $x \in G$ , we have that

$$|p_m(x) - h(x)| < \varepsilon.$$

**Example:** Projections of any real-valued function on  $G$  onto the eigenspaces of the matrix  $W$  are diffusion computable

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## Diffusion step

In a diffusion computer, computation begins with a stochastic vector, say  $\phi$  (i.e. with initial probability distribution  $p_0$  on  $V$ ).

The evolution of  $\phi$  is determined by the symmetric operator  $W$ .

*One diffusion computation step* = one application of  $W$  which maps  $\phi$  to  $W\phi$ .

A measurement is a classical inspection of the vertices of the graph  $X$  where the diffusion process takes place.

## Quantum versus diffusion steps

The principal manager of the quantum-computing group at Microsoft Research in Redmond, Washington said that

*Quantum computing is essentially matrix vector multiplication - it's linear algebra underneath the hood.*

A quantum computation entails two different types of operations:

- 1 The abstract version of the classical evolution equation in quantum mechanics, i.e. a unitary vector  $\phi$  in a Hilbert space  $(\mathbb{C}^n, \ell^2)$  evolves into a new vector  $\psi = U\phi$  ( $U$  is some unitary operator).
- 2 Measurement of this new state - quantum procedure of "collapsing  $\psi$ ": each measurement is modeled by the decomposition of  $\mathbb{C}^n$  into finite orthogonal subspaces  $H_i$ . "Collapsing" means composing  $\psi$  with the projection  $\psi_i$  onto  $H_i$ , and this projection occurs with probability  $|\psi_i|^2$ .

## Elementary considerations

$N$  a positive, odd integer which we write as a product

$$N = \prod_{i=1}^m p_i^{e_i}$$

with  $m \geq 2$  different odd prime factors with exponents  $e_i > 0$ .

Assume  $N$  is not prime nor a prime power.

To find a factor of  $N$  it suffices to find  $x$  so that

$$x^2 \equiv 1 \pmod{N} \quad \text{and} \quad x \not\equiv \pm 1 \pmod{N}. \quad (1)$$

Hint:

$$N \mid x^2 - 1 \quad \text{but} \quad N \nmid (x \pm 1).$$

# The algorithm

## Step 1.

Pick  $a \in \mathbb{Z}_N = \{1, \dots, N\}$  uniformly at random; compute  $d = \gcd(a, N)$ .

If  $1 < d < N$ , return  $d$ . Else, go to Step 2.

## Step 2.

Let  $M = \lfloor \log_2 N \rfloor + 1$  and compute the set (modular arithmetic)

$$S = \{a^{\pm 2^t} \bmod N : t = 0, \dots, M\}$$

Lemma (proved in the paper): If there are repetitions in  $S$ , then with probability  $p(m) = 1 - (m + 1)/2^m$  we can find an  $x \in \mathbb{Z}_N^*$  satisfying (1) in at most  $O(\log_2 N)$  deterministic steps.

What happens in the case when there are no repetitions in  $S$ ?

## Case no repetitions in $S$

**Step 3.** Set  $b = a^{2^M}$  and run the diffusion computer algorithm to determine the order  $r_b$  of  $b$  modulo  $N$ .

Important: the order  $r_b$  must be odd!

**Step 4.** Compute the smallest integer  $k \geq 0$  such that  $a^{2^k r_b} \equiv 1 \pmod{N}$ .

Set  $r_a = 2^k r_b$  which is the order of  $a$  modulo  $N$ .

If  $r_a$  is even, compute  $d = \gcd(a^{r_a/2} - 1, N)$ .

This algorithm produces a factor of  $N$  with probability  $1 - (m + 1)/2^m$  after  $O((\log N)^2)$  deterministic steps and some number (tbd) of diffusion steps.

## Construction of the adopted graph - multiplicative notation

Problem to be solved: Given the number  $b = a^{2^M} \in \mathbb{Z}_N^*$  determine the order  $r_b$  of  $b$  modulo  $N$  using diffusion on a suitable graph.

We know that  $r_b = r$  is odd and elements of  $S$  are distinct.

Our graph is the weighted Cayley graph  $X_{N,b} = \text{Cay}(G_{N,b}, S_{N,b}, \alpha_{N,b})$ , where

$G_{N,b} = \langle b \rangle \subseteq \mathbb{Z}_N^*$  is the subgroup of  $\mathbb{Z}_N^*$  generated by  $b$  (it has  $r$  elements)

$$S_{N,b} = \{b^{\pm 2^t} : t = 0, \dots, M\}$$

and

$$\alpha_{N,b}(b^{2^t}) := |\{l \in \{0, \dots, M\} : b^{2^t} \equiv b^{2^l} \pmod{N} \text{ or } b^{2^t} \equiv b^{-2^l} \pmod{N}\}|,$$

## Important remark

In the construction of  $X_{N,b}$  we do not know the value of  $r$ . All that is required is the value of  $N$  since we begin with one point  $b$  and, recursively, let the diffusion process develop in  $2(M + 1)$  possible directions from any given point.

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We will see that after a number of steps which is polynomial in  $\log_2 N$  we will have enough information to approximate the number of vertices of  $X_{N,b} = r$ .

What makes the process effective is the fact that diffusion occurs simultaneously at all constructed vertices, which provides some form of parallel computation.



## Construction of the adopted graph - additive notation

We replace  $X_{N,b} = \text{Cay}(G_{N,b}, S_{N,b}, \alpha_{N,b})$  with an equivalent graph ( $b$  is fixed, so omitted from notation)

$X_{r,S} = \text{Cay}(C_r, S, \alpha)$  where  $C_r = \{0, \dots, r-1\}$ ,

$$S = \{\pm 2^j : j = 0, \dots, M\} \quad \text{with} \quad M = \lfloor \log_2 N \rfloor + 1.$$

and

$$\alpha(2^j) := |\{l \in \{0, \dots, M\} : 2^j \equiv 2^l \pmod{r} \text{ or } 2^j \equiv -2^l \pmod{r}\}|.$$

$X_{r,S}$  has  $r$  vertices, and it is regular of degree  $|S| = 2(M+1)$

## Diffusion on $X_{r,S}$

Fix any point of  $X_{r,S}$ , say 0 and start diffusion with initial probability  $p_0 = (1, 0 \dots, 0)^t$ .

Recall that each step amounts to multiplying  $p_0$  by

$$W_{r,S} = \frac{1}{2} \left( 1 + \frac{1}{2(M+1)} A_{r,S} \right),$$

where  $A_{r,S}$  is the adjacency matrix of  $X_{r,S}$

We know that after  $n$  steps

$$\left| p_n^{X_{r,S}}(0) - \frac{1}{r} \right| \leq (\lambda_*^{X_{r,S}})^n, \quad (2)$$

where  $\lambda_*^{X_{r,S}}$  is the largest eigenvalue of  $W_{r,S}$  less than 1.

## Minimal number of diffusion steps to find $r$

Therefore, if we conduct  $n$  steps where  $n$  is such that  $(\lambda_*^{X_{r,s}})^n < \frac{1}{N^2}$ , the value  $p_n^{X_{r,s}}(0)$  which can be regarded as the "amount of heat" at initial point after  $n$  steps determines  $r$  uniquely through

$$r = \lfloor (p_n^{X_{r,s}}(0))^{-1} \rfloor.$$

Namely, for any two distinct positive integers  $m_1, m_2 < N$ , the smallest distance between  $1/m_1$  and  $1/m_2$  is bounded from below by

$$\frac{1}{N} - \frac{1}{N-1} = \frac{1}{N(N-1)} > \frac{1}{N^2}.$$

Thus, if we have  $p_n^{X_{r,s}}(e) = 1/r$  within an error of  $1/N^2$ , we have determined  $r$ .

# What is $\lambda_*^{X_{r,S}}$

Use known results from the graph theory:

The eigenvalues  $\eta_k$  for  $k = 0, \dots, r - 1$  of the adjacency matrix  $A_{r,S}$  are:  
 $\eta_0 = 2(M + 1)$  and

$$\eta_k = \sum_{x \in S} \alpha(x) e^{\frac{2\pi i}{r} kx} = \sum_{j=0}^M e^{\frac{2\pi i}{r} k2^j} + \sum_{j=0}^M e^{-\frac{2\pi i}{r} k2^j} \quad \text{for each } 1 \leq k \leq r - 1.$$

Therefore, eigenvalues of  $W_{r,S}$  are

$$\lambda_k^{X_{r,S}} = \frac{1}{2} \left( 1 + \frac{\eta_k}{2(M + 1)} \right) \quad \text{for } k = 0, 1, \dots, r - 1.$$

and  $\lambda_*^{X_{r,S}}$  is the largest of  $\lambda_k^{X_{r,S}}$ ,  $k = 1, \dots, r - 1$ .

# Bounding $\lambda_*^{X_{r,S}}$

One needs a bound independent of  $r$  for the quantity

$$\frac{1}{2(M+1)} \max_{k \in \{1, \dots, r-1\}} |\eta_k| = \frac{1}{2(M+1)} \max_{k \in \{1, \dots, r-1\}} \left| \sum_{j=0}^M e^{\frac{2\pi i}{r} k 2^j} + \sum_{j=0}^M e^{-\frac{2\pi i}{r} k 2^j} \right|.$$

This is a non-trivial task because the trigonometric sum is short.

Using a combination of results one arrives at the bound

$$\frac{1}{2(M+1)} \max_{k \in \{1, \dots, r-1\}} |\eta_k| < \frac{2M}{2(M+1)} = 1 - \frac{1}{M+1},$$

which yields the bound

$$\lambda_*^{X_{r,S}} \leq \frac{1}{2} \left( 1 + 1 - \frac{1}{M+1} \right) = 1 - \frac{1}{2(M+1)}.$$

## Minimal number of diffusion steps

We can now find  $n$  so that  $(\lambda_*^{X_{r,s}})^n < \frac{1}{N^2}$  by solving

$$\left(1 - \frac{1}{2(M+1)}\right)^n < \frac{1}{N^2}.$$

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Conclusion: we proved that after

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heat steps the value  $r$  equals  $\lfloor (p_n^{X_{r,s}}(0))^{-1} \rfloor$ , where  $p_n^{X_{r,s}}(0)$  is the heat at initial point.

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Note that the minimal number of quantum steps to compute the order is  $O((\log N)^2 \log \log N)$ .



## Overview of the algorithm

The algorithm runs Step 1 and either finds a factor or proceeds to Step 2.

At Step 2 we use repeated squaring, so it takes at most  $O(2M)$  steps. The algorithm can terminate at Step 2 with no answer; the probability of success if there are repetitions is at least  $1 - (m + 1)/2^m$ .

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If there are no repetitions in  $S$ , Step 3 is run on a diffusion computer and gives the answer  $r$ .

Then, Step 4 is run on a classical computer and produces a factor of  $N$  with probability at least  $1 - (m + 1)/2^m$ .

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Then, Step 4 is run on a classical computer and produces a factor of  $N$  with probability at least  $1 - (m + 1)/2^m$ .

Therefore, the algorithm terminates after at most  $O(\log N)^2$  deterministic steps plus at most  $O(\log N \log_2 N)$  diffusion steps and finds a factor of  $N$  with probability at least  $1 - (m + 1)/2^m$ .

## Example: $N = 1363$

- Step 1.** We choose  $a = 991$  which is relatively prime to 1363.
- Step 2.**  $M = \lfloor \log_2(N) \rfloor + 1 = 11$ , and  
 $S = \{991^{2^0} = 991, 991^{2^1} = 721, \dots, 991^{2^{11}} = 944, \dots\} \pmod{1363}$ . There are no repetitions in  $S$ .
- Step 3.** Set  $b = 991^{2^{11}} \equiv 944 \pmod{1363}$  and check for repetitions in the set  $S_b = \{b^{\pm 2^t} : t = 0, \dots, 11\}$  finding none. Thus, we run the diffusion computer in order to determine the order  $r_b$  of  $b = 944$ .

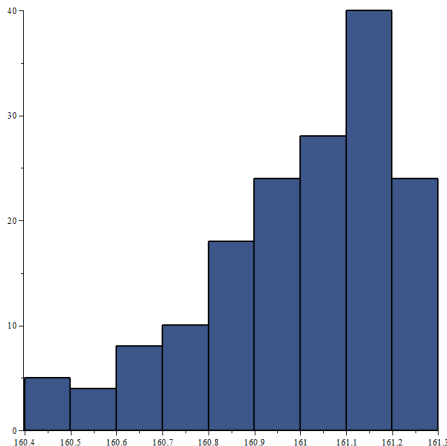
**Note:** Theoretical bound says that the diffusion computer requires at least  $\lfloor 4(M+1) \log(N) \rfloor + 1 = 347$  diffusion steps and one measurement.

Actually, after  $n = 25$  iterations, and 11 measurements we were able to conclude that  $r_b = 161$ .

- Step 5.** The smallest non negative integer  $k$  such that  $991^{2^k \times 161} \equiv 1 \pmod{1363}$  turns out to be 1. We conclude that  $r_a = 322$ .  
 Then we computed  $\gcd(991^{161} - 1, 1363) = 47$ , thus,  $N = 47 \times 29$ .

## Example: $N = 1363$ cont'

After  $n = 25$  times we measured  $p_{25}(v)$  for the 11 values of vertices  $v$  corresponding to  $S$ . For each such  $v$  we had that  $160 < p_{25}(v)^{-1} < 162$ , hence  $r_b \in \{160, 161, 162\}$ . By trying these values we confirmed that  $r_b = 161$ .



# The end

Thank you!