# Grover's algorithm

and

applications

by

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### **Plan**

- 1. Grover's algorithm
  - (a) NP
  - (b) Grover's iteration
  - (c) Search algorithm
  - (d) applications
  - (e) Heuristics
  - (f) Optimality of the algorithm
- 2. Approximate counting
  - (a) Grover's iteration + QFT = Counting
  - (b) Counting algorithm and analysis
  - (c) Severals accuracy levels
  - (d) Applications

### NP

A set S is in  $\ensuremath{\mathsf{NP}}$  if there is a polynomial time algorithm F such that

$$\forall w \in S, \exists x, F_w(x) = 1$$

$$\forall w \not\in S, \forall x, F_w(x) = 0$$

A set is **NPC** if it is in **NP** and every set in **NP** reduces to it in polynomial time.

### Example of NPC problem

#### **Scheduling:**

Given a set of constraints C find a schedule s without conflicts. Thus  $F_C(s)=1$  iff s is a schedule without conflicts in C.

#### Travelling salesman:

Given a fixed budget c and the cost to travel between a list of cities C, give a tour t with cost less than the budget c. Thus  $F_{(C,c)}(t)=1$  iff t is an appropriate tour.

#### **Knapsack:**

Given a list of objects L with their weights and values, is it possible to get a subset with value at least v and with a total weight of w. Thus  $F_{(L,v,w)}(s)=1$  iff s is appropriate.

#### Satisfiability:

Given a Boolean expression E, give an assignment to the Boolean variables  $x_i$  such that  $E(x_1, ..., x_n) = 1$ . Thus  $F_E(x) = 1$  iff E(x) = 1.

### Search Problem

#### Searching a database

Given a table T and an entry y, find i such that T[i] = y.

#### Searching under computable constraints

Given a boolean function  $F: X \to \{0, 1\}$  find x such that F(x) = 1.

**Note:** It clearly relates to **NP** problems.

### **Grover's Iteration**

$$G_F = -HS_0HS_F$$

$$S_0|i\rangle = \begin{array}{cc} -|i\rangle & \text{if } i=0 \\ |i\rangle & \text{otherwise.} \end{array}$$

$$S_F |i
angle = egin{array}{ccc} -|i
angle & ext{if } F(i) = 1 \ |i
angle & ext{otherwise.} \end{array}$$

$$H|0\rangle = \frac{1}{\sqrt{2}}(|0\rangle + |1\rangle)$$

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

$$H^{\otimes n} \ket{j} = \frac{1}{\sqrt{2^n}} \sum_{i=0}^{2^n-1} (-1)^{i \cdot j} \ket{i}$$

# Grover's Algorithm

#### Grover(F, m)

1. 
$$|\Psi\rangle \leftarrow H|0\rangle$$

2. Do m times  $|\Psi\rangle \leftarrow G_F |\Psi\rangle$ 

3. Measure  $|\Psi\rangle$  and output its value.

$$N = |X|$$
  $t = |\{x \in X | F(x) = 1\}|$ 

# **Success probability**

# Soufflé

### Iteration analysis

$$|A\rangle = \sum_{F(x)=1} |x\rangle$$
  $|B\rangle = \sum_{F(x)=0} |x\rangle$ 

$$|A\rangle + |B\rangle = \sum_{x \in X} |x\rangle$$

$$\langle A|A\rangle = t \qquad \langle B|B\rangle = N - t$$

$$H|0\rangle = \frac{1}{\sqrt{N}} \sum_{x \in X} |x\rangle = \frac{1}{\sqrt{N}} |A\rangle + \frac{1}{\sqrt{N}} |B\rangle$$

#### Solve:

$$G_F^m(H|0\rangle) = k_m|A\rangle + \ell_m|B\rangle$$

## Iteration analysis (2)

$$G_{F} |\Psi\rangle = -HS_{0}HS_{F} (k|A\rangle + \ell|B\rangle)$$

$$= HS_{0}H (k|A\rangle - \ell|B\rangle)$$

$$= H(I - 2|0\rangle\langle 0|)H (k|A\rangle - \ell|B\rangle)$$

$$= (I - \frac{2}{N}(|A\rangle + |B\rangle)(\langle A| + \langle B|)) (k|A\rangle - \ell|B\rangle)$$

$$= k|A\rangle - \ell|B\rangle + \left(-\frac{2t}{N}k + 2\frac{N-t}{N}\ell\right)(|A\rangle + |B\rangle)$$

$$= \left(\frac{N-2t}{N}k + \frac{2(N-t)}{N}\ell\right)|A\rangle$$

$$+ \left(\frac{-2t}{N}k + \frac{N-2t}{N}\ell\right)|B\rangle$$

### Iteration analysis

#### Theorem:

Let

$$\sin^2\theta = t/N$$

then

$$(G_F)^m(H|0\rangle) = k_m \sum_{F(x)=1} |x\rangle + \ell_m \sum_{F(x)=0} |x\rangle$$

where

$$k_m = \frac{\sin((2m+1)\theta)}{\sqrt{t}}$$

$$\ell_m = \frac{\cos((2m+1)\theta)}{\sqrt{N-t}}$$

### When t is known

#### Theorem:

When

$$m = \lfloor \frac{\pi}{\arcsin\left(\sqrt{t/N}\right)} \rfloor \in O(\sqrt{N/t})$$

**Grover**(F, m) outputs x such that F(x) = 1 with probability at least  $\frac{N-t}{N}$ .

#### **Proof:**

Just put the appropriate value of m in the amplitude equations of the previous slide.

### When t is unknown

#### Theorem:

There exists a quantum algorithm **Search** that given F with t>0 finds x such that F(x)=1 with expected time in  $O(\sqrt{N/t})$ .

#### Search(F)

1. 
$$m = 1, \lambda = 8/7$$

2. 
$$j \in_R \{0, \dots, m-1\}$$

3. 
$$x = Grover(F, j)$$

4. If F(x) = 1 then output x and stop

5. 
$$m = min(\lambda, \sqrt{N})$$

6. goto step 2.

**Note:** we can add a threshold of  $O(\sqrt{N})$  if we are not sure that there is a solution.

### **Minimum**

#### Theorem:

There exists an algorithm **Minimum** that finds  $x_0$  such that  $\forall x$ ,  $F(x) \geq F(x_0)$ , with probability 1/2, with an expected  $O(\sqrt{N})$  calls to F.

#### Minimum(T)

- 1.  $x_0 \in_R \{0, \dots, N-1\}$
- 2. Define F such that  $F(x) = 1 \Leftrightarrow T(x) < T(x_0)$
- 3.  $x_1 = \mathbf{Search}(F)$
- 4. If  $T(x_1) < T(x_0)$  then  $x_0 \leftarrow x_1$
- 5. If the cumulative number of calls to T is less than  $25\sqrt{N}$  goto step 2
- 6. Output  $x_0$ .

### Collision

#### Theorem:

Given  $G: X \to Y$  a two-to-one function with |X| = N, the algorithm **Collision** finds  $(x_0, x_1)$  such that  $G(x_0) = G(x_1)$  in time and space  $O(\sqrt[3]{N})$ .

#### Collision(T)

- 1. For i from 1 to  $\sqrt[3]{N}$  set T[i]=(i,G(i)).
- 2. Sort T and look for collision in T
- 3. Define  $F(x) = 1 \Leftrightarrow (x \ge \sqrt[3]{N} \text{ and } G(x) \in T)$
- 4. Set  $x_0 = \mathbf{Search}(F)$  and  $x_1$  such that  $G(x_1) = G(x_0)$
- 5. Output  $(x_0, x_1)$ .

### **Optimality**

#### Theorem:

There is no algorithm that solves the problem **Search** with good probability with an expected number of call to F less than  $\Omega(\sqrt{N})$ .

#### **Proof sketch:**

Search start in state  $|\Psi\rangle$  and call F via oracle  $O_x$ .

$$|\Psi_k^x\rangle = U_k O_x U_{k-1} \dots U_1 O_x |\Psi\rangle$$

$$|\Psi_k\rangle = U_k U_{k-1} \dots U_1 |\Psi\rangle$$

$$D_k = \sum_x ||\Psi_k^x\rangle - |\Psi_k\rangle||^2$$

#### Prove that:

- 1)  $D_K$  grows no faster than  $O(k^2)$ ,
- 2)  $D_k$  must be in  $\Omega(N)$  to distinguish N alternatives.

### **Examples of heuristics**

**Hill-Climbing:** local variations that increase an objective function. Often very efficient!

**Example:** 3-Satisfiability, find assignment to  $\{x_1, x_2, x_3, x_4\}$  that satisfies

$$(\bar{x_1} \lor \bar{x_4} \lor \bar{x_2})(\bar{x_1} \lor x_2 \lor \bar{x_3})(\bar{x_2} \lor \bar{x_4} \lor x_3)$$
  
 $(x_1 \lor \bar{x_1} \lor x_4)(x_4 \lor x_3 \lor x_3)(\bar{x_3} \lor \bar{x_4} \lor \bar{x_2})$ 

Random assignment:

$$x_1=1,\ x_2=1,\ x_3=1$$
 and  $x_4=1$  satisfies 4 clauses local variation  $x_1=0$  satisfies 5 clauses local variation  $x_2=0$  satisfies all 6 clauses!

### **Heuristics**

Let  $\mathcal{F}$  be a family of functions of the form  $F:X\to\{0,1\}$  and  $\mathcal{D}$  a probability distribution over this family.

A heuristic is a function

$$G: \mathcal{F} \times R \to X$$
.

Let 
$$t_F = |\{x|F(x) = 1\}|$$
 and  $h_F = |\{r|F(G(F,r)) = 1\}|$ 

A good heuristic is such that

$$E_{\mathcal{F}}\left(\frac{h_F}{|R|}\right) > E_{\mathcal{F}}\left(\frac{t_F}{|N|}\right)$$

### **Heuristics**

Let 
$$G'_F(r) = F(G(r, F))$$

#### **Algorithm:**

Output  $G(F, \mathbf{Search}(G'_F))$ 

#### **Analysis:**

Warning! In general

$$\left(\sum x_i\right)^{1/2} \le \sum \sqrt{x_i}$$

but

$$\sum_{F \in \mathcal{F}} \sqrt{\frac{R}{t_F}} P_F = \sum_{F \in \mathcal{F}} \sqrt{\frac{R}{t_F}} P_F \sqrt{P_F} \le$$

$$\left(\sum_{F \in \mathcal{F}} \frac{R}{t_F} P_F\right)^{1/2} \left(\sum_{F \in \mathcal{F}} P_F\right)^{1/2} = \left(\sum_{F \in \mathcal{F}} \frac{R}{t_F} P_F\right)^{1/2}$$

## **Approximate Counting**

**Counting Problem:** given  $F: X \to \{0, 1\}$  with |X| = N find  $\tilde{t}$  a good estimate of  $t = \{x | F(x) = 1\}$ .

$ t- ilde{t} $	Quantum	Classical
$O(\sqrt{t})$	$O(\sqrt{N})$	$\Omega(N)$
$\epsilon t$	$O\left(rac{1}{\epsilon}\sqrt{rac{N}{t}} ight)$	$\Omega\left(rac{N}{\epsilon^2 t} ight)$
< 1	$O(\sqrt{t(N-t)})$	$\Omega(N)$

### Counting

The amplitude is a periodic function. The period is related to t.

When m varies from 0 to P-1  $k_m$  draws r periods of a sin function.

$$k_m = \frac{\sin((2m+1)\theta)}{\sqrt{t}}$$
$$r = P\theta/\pi$$
$$\sin^2(\theta) = \frac{t}{N}$$

Use Fourier analysis to evaluate r.

### **Basics Tools**

Parameterize Grover's iteration

$$GI_F : |m\rangle \otimes |\Psi\rangle \rightarrow |m\rangle \otimes (G_F)^m |\Psi\rangle$$

Quantum Fourier Transform

$$QFT_P : |k\rangle \to \frac{1}{\sqrt{P}} \sum_{l=0}^{P-1} e^{2\pi i \frac{kl}{P}} |l\rangle \quad k \in Z_P$$

Note that:

$$QFT_P|0\rangle = \frac{1}{\sqrt{P}} \sum_{l=0}^{P-1} |l\rangle$$

## **Algorithm**

#### Count(F, P)

1. 
$$|\Psi_0\rangle \leftarrow |0\rangle H^{\otimes n} |0\rangle$$

2. 
$$|\Psi_1\rangle \leftarrow QFT_P \otimes I^{\otimes n} |\Psi_0\rangle$$

3. 
$$|\Psi_2\rangle \leftarrow GI_F |\Psi_1\rangle$$

4. 
$$|\Psi_3\rangle \leftarrow QFT_P^{-1} \otimes I^{\otimes n} |\Psi_2\rangle$$

- 5.  $\tilde{r} \leftarrow \text{measure first register of } |\Psi_3\rangle$
- 6. Output:  $\tilde{t} = N \sin^2 \frac{\tilde{r}\pi}{P}$  (and  $\tilde{r}$  if needed)

# Counting Main Theorem

#### Theorem (Counting):

For  $\tilde{t} = Count(F, P)$  then

$$|t-\tilde{t}|<rac{2\pi}{P}\sqrt{t(N-t)}+rac{\pi^2}{P^2}N$$

with probability at least  $\frac{8}{\pi^2}$ .

### **Proof**

$$|\Psi_0\rangle = \sum_{x \in X} \frac{1}{\sqrt{P}} |0\rangle |x\rangle$$

$$|\Psi_1\rangle = \sum_{m=0}^{P-1} \sum_{x \in X} \frac{1}{\sqrt{PN}} |m\rangle |x\rangle$$

$$|\Psi_2\rangle = \sum_{m=0}^{P-1} \frac{1}{\sqrt{P}} |m\rangle \left( k_m \sum_{F(x)=1} |x\rangle + \ell_m \sum_{F(x)=0} |x\rangle \right)$$

$$|\Psi_2\rangle = \sum_{F(x)=1} \left( \sum_{m=0}^{P-1} \frac{k_m}{\sqrt{P}} |m\rangle \right) |x\rangle + \sum_{F(x)=1} \left( \sum_{m=0}^{P-1} \frac{\ell_m}{\sqrt{P}} |m\rangle \right) |x\rangle$$

$$\left|\Psi_{2}^{\prime}\right\rangle = \frac{1}{\alpha}\sum_{m=0}^{P-1}\sin((2m+1)\theta)\left|m\right\rangle$$

### **Proof**

With extensive algebraic manipulation, one can show that

$$||R\rangle|^2 < 1 - \frac{8}{\pi^2},$$

thus with probability  $\frac{8}{\pi^2}$  we have

$$|\tilde{r} - r| < 1,$$
  $|\tilde{\theta} - \theta| < \frac{\pi}{P},$   $|\tilde{t} - t| < \frac{2\pi}{P} \sqrt{t(N - t)} + \frac{\pi^2}{P^2} N.$ 

### **Good Estimation**

#### Corollary 1:

Given F with N and t as defined before,  $\mathbf{Count}(F, c\sqrt{N})$  outputs  $\tilde{t}$  such that

$$|t - \tilde{t}| < \frac{2\pi}{c} \sqrt{t} + \frac{\pi^2}{c^2}$$

with probability  $\frac{8}{\pi^2}$  and requires exactly

$$c\sqrt{N}$$

evaluations of F.

#### **Proof:**

Replace P with  $c\sqrt{N}$  in counting theorem.

# Constant Factor Estimation

#### Corollary 2:

There exists an Algorithm CountRel(F,c) which output  $\tilde{t}$  such that

$$|t - \tilde{t}| < \epsilon t$$

with probability 2/3 and runs in expected time

$$O\left(\frac{1}{\epsilon}\sqrt{N/t}\right)$$
.

#### CountRel(F, c)

- 1. l = 0
- 2.  $l \leftarrow l + 1$
- 3.  $\tilde{t} \leftarrow \mathbf{Count}(F, 2^l)$
- 4. If  $\tilde{t}=0$  and  $2^l<2\sqrt{N}$  then goto step 2
- 5. Output  $\mathbf{Count}(F, \frac{200}{\epsilon}2^l)$

### **Probably Exact Counting**

#### Corollary 3:

There exists an algorithm  $\mathbf{Exact\_Count}$  that output  $\tilde{t}$  such that

$$\tilde{t} = t$$

with probability 2/3 and runs with expected time in

$$O(\sqrt{t(N-t)})$$

using only constant space.

#### $\mathsf{Exact}_\mathsf{Count}(F)$

- 1.  $\tilde{t_1} \leftarrow \mathbf{Count}(F, 50\sqrt{N})$  and  $\tilde{t_2} \leftarrow \mathbf{Count}(F, 50\sqrt{N})$
- 2.  $P \leftarrow \text{Min}(30\sqrt{\tilde{t_1}(N-\tilde{t_1})}, 30\sqrt{\tilde{t_1}(N-\tilde{t_1})})$
- 3. Output Count(F, P)

# Other Applications (Sum)

#### Corollary 4:

Let  $F: X \to Y$  with |X| = N and Y an ordered set of n-bit numbers between 0 and 1 and let

$$S = \sum_{i \in X} F(i).$$

There exists an algorithm  $\mathbf{Sum}$  that output  $\tilde{S}$  such that

$$|\tilde{S} - S| < \sqrt{S}$$

which runs in time  $O(n^2\sqrt{N})$ .

#### **Proof:**

Let  $F(i)_j$  be the jth bit of F(i).

 $\mathbf{Sum}(F, N, n)$  (Christoph Dürr 97)

- 1.  $S \leftarrow 0$
- 2. For j ranging from 1 to n

$$S \leftarrow S + 2^{j} \mathbf{Count}^{n}(F_{j}, \sqrt{N})$$

3. Output S.

# Other Applications (Selection)

#### **Approximate Selection Problem:**

Given  $F: X \to Y$  with |X| = N and k, find  $x_0$  such that if  $k' = |\{x|F(x) < F(x_0)\}|$  then  $|k - k'| < 2\pi\sqrt{k} + \pi^2$ .

Use binary search in combination with counting.

Can be solved in time  $O(\log(N)^2\sqrt{N})$ 

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