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Quantum complexity of random Boolean functions

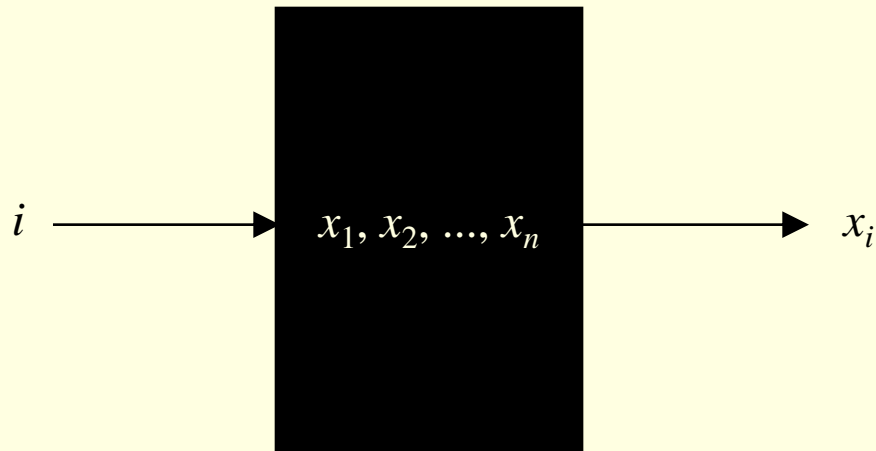
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Outline of the talk

- Query algorithm model
- Representing polynomials of Boolean functions
- Known query complexity bounds
- Our result
- Polynomial method
- Proof of our result

Query algorithm model

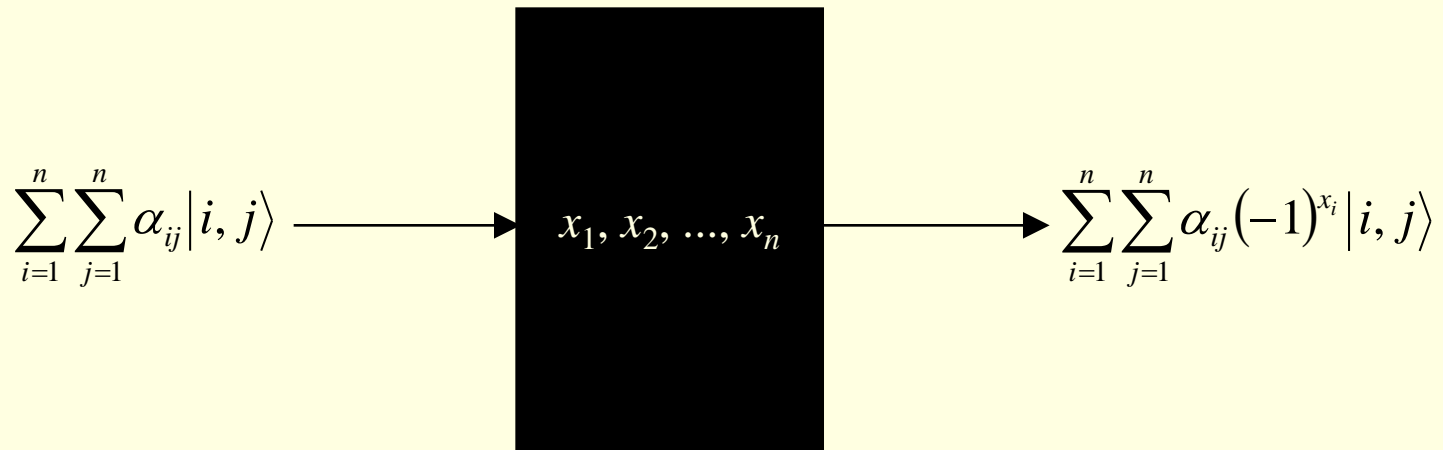
- A **known** Boolean function $f: \{0,1\}^n \rightarrow \{0,1\}$
- The input bit values are **unknown**, in a **black box** which can be **queried**, at a cost:



- The aim: to output the value of f

Query algorithm model

- **Quantum black box** query can be a superposition of the input bit indices
- Its answer is a similar superposition where each input bit is encoded in the sign of the corresponding amplitude:



- Between the queries a **quantum query algorithm** can perform any unitary transformation of its state
- In the end a 0/1 outcome measurement is performed

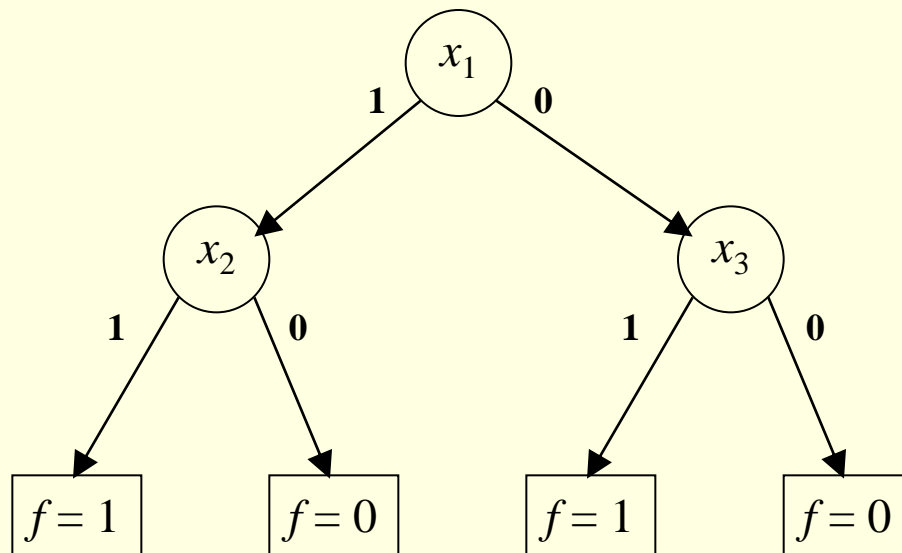
Query algorithm model

- Complexity of a query algorithm (**query complexity**): the maximum number of queries on all possible input bit strings
- Notation of the query complexity of a
 - deterministic algorithm: $D(f)$
 - quantum algorithm with success probability 1 (an **exact quantum algorithm**): $Q_E(f)$
 - quantum algorithm with error probability $\varepsilon < 1/2$: $Q_\varepsilon(f)$

Query algorithm model

- An example of a deterministic query algorithm specified as a decision tree (complexity $D(f) = 2$):

$$f(x_1, x_2, x_3) = (x_1 \wedge x_2) \vee (\neg x_1 \wedge x_3)$$



Representing polynomials of Boolean functions

- An **exact representing polynomial**: such multilinear polynomial of n variables the values of which within the domain $\{0,1\}^n$ coincide with the values of f , for example:

$$f(x_1, x_2, x_3) = (x_1 \wedge x_2) \vee (\neg x_1 \wedge x_3) = x_3 + x_1 x_2 - x_1 x_3$$

- For each Boolean function there is a unique such polynomial
- A useful identity in this domain: $a^2 = a$
- Complexity measure $deg(f)$ of a Boolean function: the degree of this polynomial

Representing polynomials of Boolean functions

- An **approximate representing polynomial**: such multilinear polynomial p the values of which are near to the values of f ($0 < \varepsilon < 1/2$):

$$\forall x_1, \dots, x_n : |p(x_1, \dots, x_n) - f(x_1, \dots, x_n)| \leq \varepsilon$$

- Approximate representing polynomial usually is not unique
- Complexity measure $deg_\varepsilon(f)$: the minimum degree of an approximate representing polynomial
- An example: the “majority function” MAJ , $\varepsilon = 1/3$,
 $deg(f) = 3$, $deg_\varepsilon(f) = 1$

$$MAJ(x_1, x_2, x_3) = x_1x_2 + x_1x_3 + x_2x_3 - 2x_1x_2x_3 \approx \frac{1}{3}x_1 + \frac{1}{3}x_2 + \frac{1}{3}x_3$$

Known query complexity bounds

- Nisan, Szegedy (1994): **for all** n -bit Boolean functions f

$$\log_2 n - O(\log \log n) \leq^* \deg(f) \leq D(f) \leq 16[\deg(f)]^8$$

* if the function f depends on all n arguments

$$\deg_\varepsilon(f) \leq \deg(f) \leq D(f) \leq c[\deg_\varepsilon(f)]^8$$

Known query complexity bounds

- Beals, Buhrman, Cleve, Mosca, de Wolf (1998): **for all** n -bit Boolean functions f

$$\frac{1}{2} \deg_{\varepsilon}(f) \leq Q_{\varepsilon}(f) \leq D(f) \leq c(\deg_{\varepsilon}(f))^6$$

and **for almost all** n -bit Boolean functions f or, equivalently, **for a random** such function f :

$$\frac{n}{2} - o(n) \leq Q_E(f)$$

(proved with the so-called polynomial method introduced in that paper)

Known query complexity bounds

- Van Dam (1998): **for all** n -bit Boolean functions f

$$Q_\varepsilon(f) \leq \frac{n}{2} + \sqrt{n}$$

- Ambainis (1999): **for almost all** n -bit Boolean functions f or **for a random** such function f with probability $1 - o(n)$:

$$\deg_\varepsilon(f) \geq \frac{n}{2} - o(n), \quad \text{therefore} \quad Q_\varepsilon(f) \geq \frac{n}{4} - o(n)$$

Open problem 1999–2012

- Where exactly between $n/4$ and $n/2$ is the bounded error query complexity Q_ε of a random Boolean function?

Our result

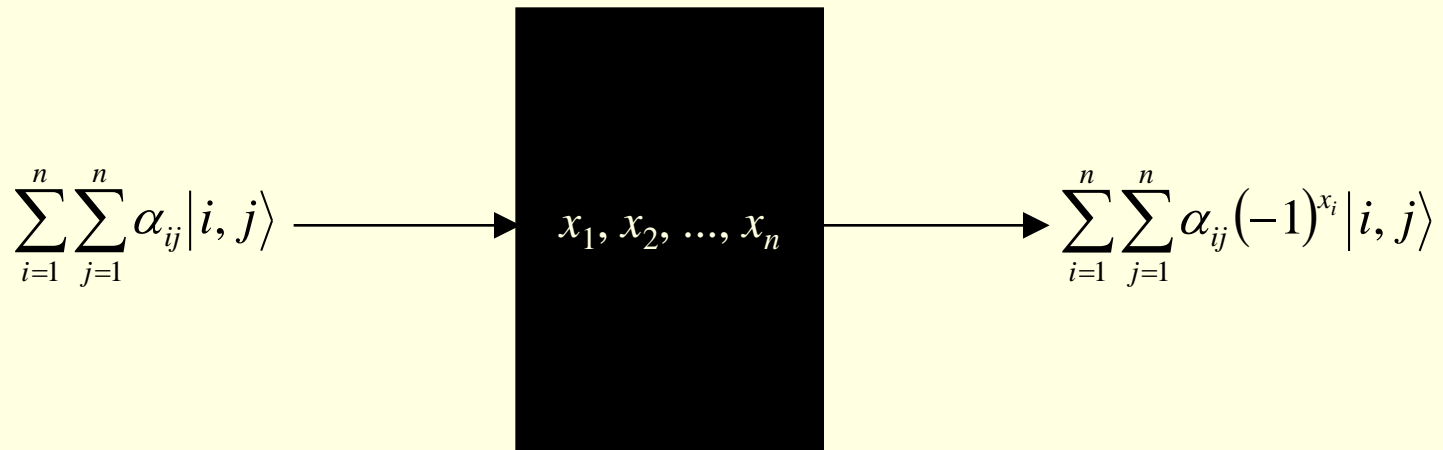
- For almost all n -bit Boolean functions f , or for a random such function f with probability $1 - o(n)$:

$$Q_\varepsilon(f) \geq \frac{n}{2} - o(n)$$

- It coincides with van Dam's upper bound up to $o(n)$ terms

Polynomial method

- **Quantum black box** query can be a superposition of the input bit indices
- Its answer is a similar superposition where each input bit is encoded in the sign of the corresponding amplitude:



- Between the queries a **quantum query algorithm** can perform any unitary transformation of its state
- In the end a 0/1 outcome measurement is performed

Polynomial method

- Beals, Buhrman, Cleve, Mosca, de Wolf (1998)
- Initially the amplitudes of a quantum query algorithm are constants
- At each query the amplitudes are multiplied by $(-1)^{x_i} = 1 - 2x_i$ (each by no more than one such factor)
- After any unitary transformation between queries each new amplitude is a linear combination of the old amplitudes
- So after d queries all amplitudes are polynomials of degree $\leq d$ of variables x_1, x_2, \dots, x_n

Polynomial method

- At the final measurement the probability p of the answer $f(x) = 1$ is the sum of modula squares of the corresponding amplitudes:

$$p(x_1, \dots, x_n) = |\alpha_1(x_1, \dots, x_n)|^2 + \dots + |\alpha_k(x_1, \dots, x_n)|^2$$

- It is thus a real polynomial of degree $\leq 2d$
- This probability must be ε -near to 1 if $f(x) = 1$ and ε -near to 0 if $f(x) = 0$ where ε is the allowed error probability; that is:

$$\forall x_1, \dots, x_n : |p(x_1, \dots, x_n) - f(x_1, \dots, x_n)| \leq \varepsilon$$

- So p is a representing polynomial, and:

$$Q_E(f) \geq \frac{1}{2} \deg(f), \quad Q_\varepsilon(f) \geq \frac{1}{2} \deg_\varepsilon(f)$$

Proof of our result

- The idea: the probability p of the answer $f(x) = 1$ is not just some arbitrary-looking approximate representing polynomial of f , it is also a sum of squares of real polynomials:

$$p(x) = (p_1(x))^2 + (p_2(x))^2 + \dots + (p_k(x))^2$$

- The result was proven by showing that the least degree of such approximate representing polynomial **which can be expressed as a sum of squares**, tends to n

Proof of our result

- If f is random, then approximately half its values are 1, and approximately half of them are 0.
- Then, since p approximates f :

$$\sum_{x:f(x)=1} p(x) \geq \approx 2^{n-1} \cdot (1 - \varepsilon) \quad \text{and} \quad \sum_{x:f(x)=0} p(x) \leq \approx 2^{n-1} \cdot \varepsilon$$

$$\text{So } \sum_{x:f(x)=1} p(x) \geq \approx \frac{1 - \varepsilon}{\varepsilon} \cdot \sum_{x:f(x)=0} p(x) \quad \text{or}$$

$$\sum_{x:f(x)=1} (p_1(x)^2 + \dots + p_k(x)^2) \geq \approx \frac{1 - \varepsilon}{\varepsilon} \cdot \sum_{x:f(x)=0} (p_1(x)^2 + \dots + p_k(x)^2)$$

$$\text{Then for at least one } j: \sum_{x:f(x)=1} (p_j(x))^2 \geq \approx \frac{1 - \varepsilon}{\varepsilon} \cdot \sum_{x:f(x)=0} (p_j(x))^2$$

Fourier representation

- Values 1 and -1 instead of the bits 0 and 1:

$$\hat{x}_i = (-1)^{x_i} = 1 - 2x_i$$

$$x_i = \frac{1 - \hat{x}_i}{2}$$

$$\hat{f}(x) = 1 - 2f(x)$$

$$\chi_s(x) = (-1)^{x_1 s_1 + x_2 s_2 + \dots + x_n s_n} = \prod_{i: s_i=1} \hat{x}_i$$

$$q_j(\hat{x}) = p_j(x) = \sum_{s: |s| \leq d} \alpha_s \chi_s(x)$$

Proof of our result

■ Then:

$$\sum_{x:f(x)=1} (q_j(\hat{x}))^2 \geq \frac{1-\varepsilon}{\varepsilon} \cdot \sum_{x:f(x)=0} (q_j(\hat{x}))^2 \Leftrightarrow$$

$$\Leftrightarrow \sum_x (q_j(\hat{x}))^2 \left(\frac{1-\hat{f}(x)}{2} \right) \geq \frac{1-\varepsilon}{\varepsilon} \cdot \sum_x (q_j(\hat{x}))^2 \left(\frac{1+\hat{f}(x)}{2} \right) \Leftrightarrow$$

$$\Leftrightarrow -\sum_x (q_j(\hat{x}))^2 \hat{f}(x) \geq (1-2\varepsilon) \cdot \sum_x (q_j(\hat{x}))^2 \Leftrightarrow$$

$$\Leftrightarrow -\sum_x \sum_{s:|s|\leq d} \sum_{s':|s'|\leq d} \alpha_s \alpha_{s'} \chi_{s\oplus s'}(x) \hat{f}(x) \geq (1-2\varepsilon) \cdot \sum_x \sum_{s:|s|\leq d} \sum_{s':|s'|\leq d} \alpha_s \alpha_{s'} \chi_{s\oplus s'}(x) \Leftrightarrow$$

$$\Leftrightarrow -\sum_{s:|s|\leq d} \sum_{s':|s'|\leq d} \alpha_s \alpha_{s'} \sum_x \chi_{s\oplus s'}(x) \hat{f}(x) \geq (1-2\varepsilon) \cdot \sum_{s:|s|\leq d} \sum_{s':|s'|\leq d} \alpha_s \alpha_{s'} \sum_x \chi_{s\oplus s'}(x) \Leftrightarrow$$

Proof of our result

$$\Leftrightarrow - \sum_{s:|s|\leq d} \sum_{s':|s'|\leq d} \alpha_s \alpha_{s'} \sum_x \chi_{s\oplus s'}(x) \hat{f}(x) \geq (1-2\varepsilon) \cdot \sum_{s:|s|\leq d} \sum_{s':|s'|\leq d} \alpha_s \alpha_{s'} \sum_x \chi_{s\oplus s'}(x) \Leftrightarrow$$

$$\Leftrightarrow - \sum_{s:|s|\leq d} \sum_{s':|s'|\leq d} \alpha_s \alpha_{s'} \left(\chi_{s\oplus s'}, \hat{f} \right) \geq (1-2\varepsilon) \cdot \sum_{s:|s|\leq d} \alpha_s^2 \quad \text{where } (g, h) = \frac{1}{2^n} \sum_x g(x)h(x)$$

$$\Leftrightarrow q_j^T \cdot \hat{F}_d \cdot q_j \geq (1-2\varepsilon) \cdot q_j^T \cdot q_j \quad \text{where } \left[\hat{F}_d \right]_{s:|s|\leq d, s':|s'|\leq d} = - \left(\chi_{s\oplus s'}, \hat{f} \right)$$

$$\Leftrightarrow \left\langle q_j \middle| \hat{F}_d \middle| q_j \right\rangle \geq (1-2\varepsilon) \cdot \left\langle q_j \middle| q_j \right\rangle$$

It follows that the largest eigenvalue of \hat{F}_d must be at least $(1-2\varepsilon)$

Actually, it tends to 0, unless $d \geq n/2 - o(n)$.

Proof of our result

- Particularly, we proved that with probability $1 - o(n)$:

$$\|\hat{F}_d\|_\infty = O\left(\sqrt{\frac{nB^{1+o(1)}}{2^n}}\right) \quad \text{where} \quad B = \sum_{i=0}^d \binom{n}{i}$$

$$\text{Since } \|\hat{F}_d\|_\infty \geq 1 - 2\varepsilon, \quad \text{we get } B \geq 2^{n-o(n)}$$

It is well known that $B \leq 2^{nH(d/n)}$ where H is the binary entropy function

So $2^{nH(d/n)} \geq 2^{n-o(n)}$ which implies $d \geq n/2 - o(n)$.

Proof of our result

- The estimation of the largest eigenvalue:

$$\mathbf{E}_f \|\hat{F}_d\|_\infty = \mathbf{E}_f \max_i |\lambda_i| \leq \sqrt[2k]{\mathbf{E}_f \max_i \lambda_i^{2k}} \leq \sqrt[2k]{\mathbf{E}_f \sum_i \lambda_i^{2k}} = \sqrt[2k]{\mathbf{E}_f \text{Tr}(\hat{F}_d^{2k})}$$

Estimating $\mathbf{E}_f \text{Tr}(\hat{F}_d^{2k})$ by writing out $\text{Tr}(\hat{F}_d^{2k})$ explicitly and algebraically manipulating it leads to the result

Open problem 1998–2014

- Where exactly between $n/2$ and n is the **exact** query complexity Q_E of a random Boolean function?

Thank you for attention!

Questions?