Polynomials, quantum query complexity, and Grothendieck's inequality

Scott Aaronson¹, Andris Ambainis², Jānis Iraids², <u>Martins Kokainis²</u>, Juris Smotrovs²

¹Computer Science and Artificial Intelligence Laboratory, MIT

²Faculty of Computing, University of Latvia

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Query model

- Function $f(x_1, x_2, ..., x_n), x_i \in \{0, 1\}.$
- x_i given by a black box:

$$i \longrightarrow x_i$$

• Complexity = number of queries.

Quantum query model



- U_0 , U_1 , ..., U_T , independent of x_1 , ..., x_n .
- O_X query operators:

$$\sum_{i} a_{i} \ket{i} \xrightarrow{O_{X}} \sum_{i} a_{i} (-1)^{x_{i}} \ket{i}$$

• $Q_{\epsilon}(f)$ – minimum number of queries in a quantum algorithm computing f correctly with probability $\geq 1 - \epsilon$.



- Lower bounds on quantum query complexity
 - OR: no polynomial of degree $o(\sqrt{n})$ approximating OR [NS94], thus no quantum algorithm making $o(\sqrt{n})$ queries.
 - Collision problem, element distinctness problem, ...
- The obtained bounds can be asymptotically lower than $Q_{\epsilon}(f)$.

Multilinear polynomials of degree *d*

 \implies [BBCMW01]

Quantum algorithms that make $O(d^6)$ queries

A multilinear polynomial of degree *d*

& [ABK16] Quantum algorithms make $\tilde{\Omega}(d^4)$ queries

Quantum algorithms that make T queries

??

Multilinear polynomials of degree 2*T*

Quantum algorithms that make T queries



Multilinear polynomials of degree 2*T*

This work:

Quantum algorithms that make 1 query



Multilinear polynomials of degree 2

- Recently shown [AA15]:
 - A task that requires 1 query quantumly and $\Theta(\sqrt{n})$ queries classically.
 - Any quantum algorithm which makes 1 query can be simulated by a probabilistic algorithm making $O(\sqrt{n})$ queries.

Multilinear polynomials

A multilinear polynomial $p : \mathbb{R}^n \to \mathbb{R}$ represents $f : (X \subset \{-1, 1\}^n) \to \{0, 1\}$ with error $\delta \in [0; 0.5)$ if

•
$$x \in X, f(x) = 0 \Rightarrow p(x) \in [0; \delta];$$

• $x \in X, f(x) = 1 \implies p(x) \in [1 - \delta; 1];$

• $p(x) \in [0; 1]$ for all $x \in \{-1, 1\}^n$.

A block-multilinear polynomial $q: \mathbb{R}^{d(n+1)} \to \mathbb{R}$ of degree d

$$q(x^{(1)},\ldots,x^{(d)}) = \sum_{i_1,i_2,\ldots,i_d=0\ldots n} a_{i_1i_2\ldots i_d} x^{(1)}_{i_1} x^{(2)}_{i_2} \ldots x^{(d)}_{i_d}, \quad x^{(j)} \in \mathbb{R}^{n+1},$$

represents $f: (X \subset \{-1,1\}^n) \rightarrow \{0,1\}$ with error $\delta \in [0;0.5)$ if

•
$$x \in X, f(x) = 0 \Rightarrow q(\tilde{x}, \tilde{x}, \dots, \tilde{x}) \in [0; \delta], \quad \tilde{x} := (1, x);$$

•
$$x \in X, f(x) = 1 \Rightarrow q(\tilde{x}, \tilde{x}, \dots, \tilde{x}) \in [1 - \delta; 1], \quad \tilde{x} := (1, x);$$

•
$$q(x^{(1)}, \ldots, x^{(d)}) \in [-1; 1]$$
 for all $x^{(1)}, \ldots, x^{(d)} \in \{-1, 1\}^{n+1}$.

Example

- Consider $NAE(x_1, x_2, x_3) = \neg (x_1 = x_2 = x_3).$
- Ordinary exact representation:

$$p(x_1, x_2, x_3) = \frac{3 - x_1 x_2 - x_1 x_3 - x_2 x_3}{4}$$

Block-multilinear exact representation:

$$q(x_0,\ldots,x_3,y_0,\ldots,y_3)=\frac{2x_0y_0-x_1y_2-x_1y_3-x_3y_2+x_3y_3}{4}$$

• Notice that setting $x_0 = y_0 = 1$ and $x_i = y_i$ yields

$$q(1, x_1, x_2, x_3, 1, x_1, x_2, x_3) = p(x_1, x_2, x_3).$$

From quantum algorithms to polynomials

- deg_ε(f): the minimum degree of a polynomial p representing f with error ε;
- bmdeg_ε(f): the minimum degree of a block-multilinear polynomial q representing f with error ε.

Theorem ([BBCMW01])

$$\mathsf{Q}_{\epsilon}(f) \geq 2\widetilde{\mathsf{deg}}_{\epsilon}(f)$$

Theorem ([AA15])

$$Q_{\epsilon}(f) \geq 2\widetilde{bmdeg}_{\epsilon}(f)$$

Theorem

$Q_{\epsilon}(f) = 1 ext{ for some } \epsilon < 0.5 \quad \Leftrightarrow \quad \widetilde{\deg}_{\delta}(f) = 2 ext{ for some } \delta < 0.5$

Sketch of the proof

- From a multilinear polynomial p to a block-multilinear polynomial q.
- **2** By splitting variables from q to a block-multilinear polynomial q'.
- **③** A quantum algorithm which estimates q' by making a single query.

Estimating a polynomial with a quantum algorithm

• A block-multilinear polynomial q of degree 2:

$$q(x_1,\ldots,x_n,y_1,\ldots,y_n)=\sum_{i=1}^n\sum_{j=1}^na_{ij}x_iy_j.$$

• Let
$$A = (a_{ij})$$
 and suppose $U = n \cdot A$ is unitary.

• One can prepare with a single query each of the states

$$|\Psi_x\rangle = \frac{1}{\sqrt{n}}\sum_{i=1}^n x_i |i\rangle, \quad |\Psi_y\rangle = \frac{1}{\sqrt{n}}\sum_{j=1}^n y_j |j\rangle,$$

thus with a single query it is possible to estimate

$$\langle \Psi_x | U | \Psi_y \rangle = q(x_1, \ldots, x_n, y_1, \ldots, y_n).$$

• Still works if $||U|| \leq C$.

Preprocessing a block-multilinear polynomial

• Have: $|q| \leq 1$, i.e.,

$$\max_{x,y\in\{-1,1\}^n} \left| \sum_{i=1}^n \sum_{j=1}^n a_{ij} x_i y_j \right| \le 1 \quad \text{ or } \quad \left\|A\right\|_{\infty \to 1} \le 1.$$

- Need: $n ||A|| \leq C$.
- Solution: variable splitting.
- A variable x_i can be replaced by new variables x_{i_1}, \ldots, x_{i_k} as follows:

$$x_i \longrightarrow \frac{x_{i_1} + x_{i_2} + \ldots + x_{i_k}}{k}$$

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- Another block-multilinear polynomial q' is obtained with a coefficient matrix A' of size n' × m'.
- $\bullet \ {\sf Still} \ |q'| \leq 1 \ {\rm or} \ \|{\it A}'\|_{\infty \rightarrow 1} \leq 1.$
- Can we achieve $\sqrt{n'm'} \|A'\| \le C$?

 Another block-multilinear polynomial q' is obtained with a coefficient matrix A' of size n' × m'.

• Still
$$|q'| \leq 1$$
 or $||A'||_{\infty \to 1} \leq 1$.

• Can we achieve
$$\sqrt{n'm'} \|A'\| \le C$$
?

Claim

For each $\delta > 0$ it is possible to split variables so that the obtained matrix A' satisfies

$$\sqrt{n'm'} \left\| \mathsf{A}' \right\| \le \mathsf{K} + \delta,$$

where K < 1.7823 – Groethendieck's constant.

Key idea: splitting variables is equivalent to factorizing the matrix A.

Splitting variables \equiv splitting rows/columns of A

• Splitting a variable x_i into k new variables corresponds to splitting the *i*th row of A into k equal rows.

Example

- Let $q = \frac{1}{2} (x_1 y_1 + x_2 y_1 + x_1 y_2 x_2 y_2)$, then $A = \begin{pmatrix} 0.5 & 0.5 \\ 0.5 & -0.5 \end{pmatrix}$.
- Replacing x_2 with $\frac{x_2'+x_3'+x_4'}{3}$ corresponds to ...

• . . . replacing A with

$$A' = \begin{pmatrix} \frac{1}{2} & \frac{1}{2} \\ \frac{1}{6} & -\frac{1}{6} \\ \frac{1}{6} & -\frac{1}{6} \\ \frac{1}{6} & -\frac{1}{6} \end{pmatrix}$$

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Example

• Let
$$q = \frac{1}{2} (x_1 y_1 + x_2 y_1 + x_1 y_2 - x_2 y_2)$$
, then $A = \begin{pmatrix} 0.5 & 0.5 \\ 0.5 & -0.5 \end{pmatrix}$

• Replacing x_2 with $\frac{x_2+x_3+x_4}{3}$ corresponds to ...

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- Suppose that A is of size $n \times m$ and its
 - 1st row is split into k_1 rows,
 - 2nd row into k_2 rows,
 - *n*th row into k_n rows,

obtaining A' of size $n' \times m'$.

- Clearly, m' = m, $n' = k_1 + k_2 + \ldots + k_n$.
- What about ||A'||?

. . .

• We have ||A'|| = ||B||, where

$$B = \begin{pmatrix} \frac{a_{11}}{\sqrt{k_1}} & \frac{a_{12}}{\sqrt{k_1}} & \cdots & \frac{a_{1m}}{\sqrt{k_1}} \\ \frac{a_{21}}{\sqrt{k_2}} & \frac{a_{22}}{\sqrt{k_2}} & \cdots & \frac{a_{2m}}{\sqrt{k_2}} \\ & & \ddots & \\ \frac{a_{n1}}{\sqrt{k_n}} & \frac{a_{n2}}{\sqrt{k_n}} & \cdots & \frac{a_{nm}}{\sqrt{k_n}} \end{pmatrix}$$

• Consequently,

$$\|A'\| \sqrt{n'm'} = \|B\| \|w\| \|v\|,$$

where $w = (\sqrt{k_1}, \dots, \sqrt{k_n}), v = (1, \dots, 1).$

Splitting rows/columns \equiv factorizing A

 \Leftrightarrow

• Let A be of size $n \times m$ and C > 0.

• Claim:

 $\exists B \in \mathbb{R}^{n \times m}$ and $w \in \mathbb{R}^n_+$, $v \in \mathbb{R}^m_+$:

- $a_{ij} = w_i b_{ij} v_j, \quad \forall i, j,$
- $w_i^2, v_j^2 \in \mathbb{Q}, \forall i, j$,

• ||B|| ||w|| ||v|| = C

 $\exists A' \in \mathbb{R}^{n' \times m'}:$ • $A \longrightarrow A',$ • $\|A'\| \sqrt{n'm'} = C$

Splitting rows/columns \equiv factorizing A

• Let A be of size $n \times m$ and C > 0.

• Claim:

 $\exists B \in \mathbb{R}^{n \times m}$ and $w \in \mathbb{R}^n_+$, $v \in \mathbb{R}^m_+$:

- $a_{ij} = w_i b_{ij} v_j, \quad \forall i, j,$
- $w_i^2, v_j^2 \in \mathbb{Q}, \forall i, j,$
- ||B|| ||w|| ||v|| = C

 $\forall \delta > 0 \; \exists A' \in \mathbb{R}^{n' \times m'}:$ • $A \longrightarrow A',$ • $\|A'\| \sqrt{n'm'} = C + \delta$

Grothendieck's Inequality: I

Suppose that

- A is a $n \times m$ matrix with real components;
- \mathcal{H} is an arbitrary Hilbert space;
- $\mathbf{x}_1, \ldots, \mathbf{x}_n, \mathbf{y}_1, \ldots, \mathbf{y}_m \in \mathcal{H}$ are of norm at most 1.

Then

$$\left|\sum_{i=1}^{n}\sum_{j=1}^{m}a_{ij}\left\langle \mathbf{x}_{i},\mathbf{y}_{j}\right\rangle \right|\leq K\left\|A\right\|_{\infty\rightarrow1},$$

where

$$\|A\|_{\infty \to 1} = \max_{\substack{x \in \{-1,1\}^n \\ y \in \{-1,1\}^m}} \left| \sum_{i=1}^n \sum_{j=1}^m a_{ij} x_i y_j \right|.$$

Grothendieck's Inequality: II

- Suppose that A is a $n \times n$ matrix. Then the following are equivalent:
 - for each \mathcal{H} and all \mathbf{x}_i , $\mathbf{y}_j \in \mathcal{H}$ (of norm ≤ 1), $i, j \in [n]$,

$$\left|\sum_{i=1}^{n}\sum_{j=1}^{n}\mathsf{a}_{ij}\left\langle \mathsf{x}_{i},\mathsf{y}_{j}
ight
angle
ight|\leq1;$$

2 there is an $n \times n$ matrix B and vectors $w, v \in \mathbb{R}^n_+$, s.t.

•
$$||w|| = ||v|| = 1;$$

- $||B|| \le 1;$
- $w_i b_{ij} v_j = a_{ij}$ for all i, j.

Putting everything together

• Since $||A||_{\infty \to 1} \leq 1$, there is a matrix *B* and vectors *w*, *v* s.t.

 $||w|| = ||v|| = 1, ||B|| \le K$ and $w_i b_{ij} v_j = a_{ij}$ for all i, j.

- Then we can split variables so that the obtained matrix A' satisfies $||A'|| \sqrt{n'm'} \le K + \delta$, for every $\delta > 0$.
- Therefore there is a 1-query quantum algorithm which estimates q' (the polynomial corresponding to A'),
- thus evaluating the polynomial q.

 $deg = 2 \Rightarrow bmdeg = 2$

Claim

Suppose that

- $p : \mathbb{R}^n \to \mathbb{R}$ is a multilinear polynomial of degree 2,
- $|p(x)| \le 1$ for each $x \in \{-1, 1\}^n$.

Then there exists a block-multilinear polynomial $g : \mathbb{R}^{2n+2} \to \mathbb{R}$ s.t.

- deg g = 2, • $g(\tilde{x}, \tilde{x}) = \frac{1}{3}p(x)$, $\tilde{x} := (1, x)$, for each $x \in \{-1, 1\}^n$,
- $|g(z)| \le 1$ for each $z \in \{-1, 1\}^{2n+2}$.

From polynomials to block-multilinear polynomials

Claim

Suppose that

- $p: \mathbb{R}^n \to \mathbb{R}$ is a multilinear polynomial of degree d,
- $|p(x)| \le 1$ for each $x \in \{-1, 1\}^n$.

Then there exists a block-multilinear polynomial $g: \mathbb{R}^{d(n+1)} \to \mathbb{R}$ s.t.

- deg g = d, • $g(\tilde{x}, ..., \tilde{x}) = p(x)$ for each $x \in \{-1, 1\}^n$, $\tilde{x} := (1, x)$;
- $|g(z)| \le C_d = O(3.5911...^d)$ for each $z \in \{-1, 1\}^{d(n+1)}$.

Key ideas:

 replace each monomial with its symmetric block-multilinear version (average over all the ways how one could use one term per block), e.g.,

$$x_1 x_2 \dots x_r \longrightarrow \frac{1}{\binom{d}{r} r!} \sum_{\substack{B \subset [d]: \\ |B| = r}} \sum_{\substack{b: \\ b - \text{bijection}}} x_1^{(b(1))} x_2^{(b(2))} \dots x_r^{(b(r))}$$

2 Apply the polarization identity to show the boundedness of g:

$$d!F\left(u^{(1)},u^{(2)},\ldots,u^{(d)}\right) = \sum_{\substack{\mathcal{T}\subset [d]\\\mathcal{T}\neq\emptyset}} (-1)^{d-|\mathcal{T}|} f\left(\sum_{\substack{j\in\mathcal{T}}} u^{(j)}\right),$$

where f(x) := F(x, x, ..., x) and $F : E^d \to \mathbb{R}$ is a *d*-linear and symmetric map.

• Corollary: solution of an open problem from [AA15].

Claim

Let $g : \mathbb{R}^n \to \mathbb{R}$ be a multilinear polynomial of degree d with $|g(y)| \leq 1$ for any $y \in \{-1,1\}^n$. Then g(y) can be approximated within precision $\pm \epsilon$ whp by querying $O((\frac{n}{\epsilon^2})^{1-1/d}))$ variables (with a big-O constant depending on d).

• The same result (and transformation of ordinary multilinear polynomials to block-multilinear ones) has been independently shown by O'Donnell and Zhao by means of decoupling theory.

• Q and bmdeg are not equivalent: there is a function exhibiting a quadratic separation between both measures.

Theorem

There exists f with $\mathsf{Q}_{\epsilon}(f) = \tilde{\Omega}(\mathsf{bmdeg}_0^2(f))$.

- Recently [ABK16] an analogous result for Q_{ϵ} and deg₀ using the cheat sheet framework.
- We show that the same function provides the separation between Q_{ϵ} and $\mathsf{bmdeg}_0.$

? Characterize quantum algorithms with 2, 3, ..., queries?

? 2 queries \equiv polynomials of degree 4?

Thank you for your attention!

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