Quantum Query Complexity of Boolean Functions with Small On-Sets

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Motivation

 Want to test some properties of huge data X, Or, compute some function f(X).

- e.g. WWW log analysis, Experimental data analysis....



- Reading all memory cells of X costs too much.
- Can we save the number of accessing X when computing certain functions f(X) ?

Oracle Computation Model

Can know the value of one cell by making a query to X.



- Cost measure:= # of queries to be made.
 (All other computation is free.)
- R(f): Query complexity of f

:= # of queries needed to compute f for the worst input X

Oracle Computation Model

 Can know the value of one cell by making a query to X.





Quantum Computation

Qubit: A unit of quantum information.

A quantum state $|\phi\rangle$ of one qubit : a unit vector in 2-dimensional Hilbert space.

For an orthonormal basis $\left(\begin{pmatrix} 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \end{pmatrix} \right) = \left(|0\rangle, |1\rangle \right)$, $|\phi\rangle = \alpha |0\rangle + \beta |1\rangle$ where $\alpha, \beta \in \mathbb{C}$ and $|\alpha|^2 + |\beta|^2 = 1$.

A quantum state $|\varphi\rangle$ of n qubits : a unit vector in 2^n - dimensional Hilbert space.

$$|\varphi\rangle = \sum_{i=0}^{2^n-1} \alpha_i |i\rangle$$
 for orthonormal basis $\{i\}_i$.

Quantum operation: only unitary operation $H|\varphi\rangle \rightarrow |\varphi'\rangle$

Oracle Computation Model (Quantum)

- A quantum query is a linear combination of classical queries.
- Can know a linear combination of the value of cells per query.



- Q(f): (Bounded-error) Quantum query complexity of f
 - := # of quantum queries needed to compute f with error probability < 1/3 for the worst input X

Fundamental Problems

- What is the quantum/classical query complexity of function f?
- For what function f, is quantum computation faster than classical one?

In particular, Boolean functions are major targets.

This talk focuses on Boolean functions in bounded-error setting (constant error probability is allowed).

Previous Works

- (Almost) No quantum speed up against classical.
 PARITY, MAJORITY [BBCMdW01].
 - $\Omega(N)$ quantum queries are needed.
- Polynomial quantum speed up against classical
 - OR [Gro96], AND-OR trees [HMW03, ACRSZ07]
 - Quantum O(\sqrt{N}) v.s. Classical Ω (N).
 - k-threshold functions for k<< N/2 [BBCMdW01]
 - Quantum $\Theta(\sqrt{(kN)})$ v.s. Classical Ω (N).
 - Testing graph properties (N=n(n-1)/2 variables)
 - Triangle: Quantum O(n^{1.3}) [MSS05]
 - Star: Quantum $\Theta(n^{1.5})$ [BCdWZ99]
 - Connectivity: Quantum $\Theta(n^{1.5})$ [DHHM06]

But much less is known except for the above typical cases.

 \rightarrow We investigate the query complexity of the families defined a natural parameter.

Classical $\Omega(n^2)$

On-set of Boolean Functions

We consider the *size of the on-set* of a Boolean function as a parameter.

On-set S_f of a Boolean function f: The set of input X \in {0,1}^N for which f(X)=1.

Ex.) On-set S_f of $f=(x_1 \land x_2) \lor x_3$: $(x_1,x_2,x_3)=(1,1,0), (1,1,1), (0,0,1), (0,1,1), (1,0,1).$

The size of S_f is 5.

Our Results (1/2)



Our results (2/2)

Our hardest-case complexity gives the tight complexity of some graph property testing.



• (Planarity testing) Is G planar? : $Q(f)=\Theta(n^{1.5})$. $R(f)=\Omega(n^2)$

(For a given adjacency list, O(n) time complexity [Hopcroft-Tarjan74])

•(Graph Isomorphism testing) Is G isomorphic to a fixed graph G'?:

 $Q(f) = \Theta(n^{1.5}).$ (R(f) = $\Omega(n^2)$ [DHHM06])

By setting M = # of graphs with property P.

OUTLINES OF PROOFS

Our Results(1/2)



Hardest-case Bound

Theorem : For any function $f \in F_{N,M}$, $Q(f) = \Theta\left(\sqrt{N \frac{\log M}{\log N}}\right)$ if $poly(N) \le M \le 2^{N^d}$ for some constant d(0 < d < 1).

Proof.

Lower Bound: By showing a function for every *M* which has $O\left(\sqrt{N\frac{\log M}{\log N}}\right)$ complexity. (The function is similar to *t* - threshold function for $t = \frac{\log M}{\log N}$.)

Upper bound:

Use the algorithm [AIKMRY07] for Oracle Identification Problem.

Oracle Identification Problem (OIP)

 Given a set of M candidates, identify the N-bit string in the oracle.

Oracle (N	[=8)											
i	0	1	2	3	4	5	6	7				
X _i	?	?	?	?	?	?	?	?				
Candidate Set (N=8, M=4) Can see the contents w/o making queries												
i	0	1	2	3	4	5	6	7				

l	U		Ζ	3	4	C	Ö	1
Candidate 1	0	1	1	1	0	0	0	0
Candidate 2	1	1	0	1	0	1	1	0
Candidate 3	1	0	1	0	1	0	0	0
Candidate 4	0	0	0	1	1	0	0	0

Hardest-case Bound

Proof (Continued)



Idea:

- •Set the onset S_f to the candidate set of OIP and run the algorithm for OIP to get an estimate $Y \in S_f$ of X.
- •By definition, Y=X (with high probability) iff f(X)=1.

Test if X=Y, which can be done with quantum query complexity $O(\sqrt{N})$.

Our Results (1/2)



Easiest-case Bound

Theorem : If $M \le 2^{\frac{N}{2+\varepsilon}}$ for any positive constant ε , $Q(f) = \Theta(\sqrt{N})$ for any $f \in F_{N,M}$.

Proof: Use sensitivity argument.

Th.[Beals et al. 2001] $Q(f) = \Omega(\sqrt{s(f)})$

Assuming s(f)=o(N), we can conclude a contradiction by simply counting, $|f^{-1}(1)| > 2^{\frac{N}{2+\varepsilon}} \ge M$

We can construct a function with such quantum query complexity.

Our Results (1/2)



Theorem : Average of Q(f) over all $f \in F_{N,M}$ is $O(\log M + \sqrt{N})$.

Proof.

Claim: For almost all functions f in $F_{N,M}$, every element in the on-set S_f differs from any other in the first O(log M) bits.

O(log M) bits. Y_1 1011.....01 Y_{M-1} 1101.....01 Y_M 0001.....00

- Make queries to the first O(log M) bits to identify a unique string Y in S_f (If there is no such Y, we are done: f(X)=0.)
- 2. Test if Y=X with O(\sqrt{N}) quantum queries. Y=X if and only if f(X)=1.

Theorem : Average of
$$Q(f)$$
 over all $f \in F_{N,M}$ is $O\left(\frac{\log M}{c + \log N - \log \log M} + \sqrt{N}\right)$.

Proof

With one quantum query,
$$|\varphi_X\rangle = \frac{1}{\sqrt{N}} \sum_{i=1}^{N} (-1)^{x_i} |i\rangle$$
.

Claim: For almost all functions in $F_{N,M}$, every X,Y in the onset S_f satisfy $\left|\left\langle \varphi_X \left| \varphi_Y \right\rangle\right| = \left|\frac{1}{N} \left(N - 2Ham(X,Y)\right)\right| > 2\sqrt{\frac{\log M}{N}}.$

(Proof is by bounding Hamming distance with coding-theory argument and Chernoff-like bound.)

$$\langle \varphi_X | \varphi_Y \rangle$$
 is large enough to identify X in S_f with

$$O\left(\frac{\log M}{c + \log N - \log \log M}\right) \text{copies of } |\varphi_{\mathsf{X}}\rangle$$

according to quantum state discrimination theorem [HW06].

Theorem : Average of Q(f) over all $f \in F_{N,M}$ is $\Omega(\log M / \log N + \sqrt{N}).$

Actually, we prove stronger statement.

Theorem : Average of unbounded - error query complexity over all $f \in F_{N,M}$ is $\Omega(\log M / \log N + \sqrt{N})$.

Unbounded-error: error probability is 1/2- ϵ for arbitrary small ϵ

Proof: Use the next theorem.

Theorem[Anthony1995 + Next Talk] The number of Boolean functions f whose unbounded query complexity is d/2 is

$$T(N,d) \le 2\sum_{k=0}^{D-1} \binom{2^N - 1}{k} \text{ for } D = \sum_{i=0}^d \binom{N}{i}.$$

For
$$d = \frac{\log M}{2\log N}$$
, we can prove

$$T\left(N, \frac{\log M}{2\log N}\right)$$
 is much smaller than $\binom{2^N}{M}$, i.e., the size of $F_{N,M}$.



Application: Planarity Testing



we carefully prepare a set of planar graphs and a set of non-planar graphs ,

and then apply the quantum/classical adversary method [Amb01,Aar04].

Summary

- Proved the tight quantum query complexity of the family of Boolean functions with fixed on-set size M.
- Functions with on-set size M have various quantum query complexity, while their randomized query complexity is $\Omega(N)$ for $poly(N) \le M \le 2^{N^d}$.

(For large M, the functions may have small randomized query complexity.)

- On-set size is a very simple and natural parameter, which enables us to easily analyze the query complexity of some Boolean functions with our bounds.
- In particular, we proved the tight quantum query complexity of some graph property testing problems.

Thank you!