

# Query Complexity: Worst-Case Quantum Versus Average-Case Classical

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## 1 Introduction

In this note we investigate the relationship between worst-case quantum query complexity and average-case classical query complexity. Specifically, we show that if a quantum computer can evaluate a total Boolean function  $f$  (with bounded error) using  $T$  queries in the worst case, then a deterministic classical computer can evaluate  $f$  using  $O(T^5)$  queries in the average case, under a uniform distribution of inputs. (If  $f$  is monotone, we show furthermore that only  $O(T^3)$  queries are needed.)

Previously, Beals et al. [3] showed that if a quantum computer can evaluate  $f$  (with bounded error) using  $T$  queries in the worst case, then a deterministic classical computer can evaluate  $f$  using  $O(T^6)$  queries in the worst case, or  $O(T^4)$  if  $f$  is monotone. The optimal bound is conjectured to be  $O(T^2)$ , but improving on  $O(T^6)$  remains an open problem. Relating worst-case quantum complexity to average-case classical complexity may suggest new ways to reduce the polynomial gap in the ordinary worst-case versus worst-case setting.

## 2 Preliminaries

Let  $f : \{0,1\}^n \rightarrow \{0,1\}$  be a total Boolean function. Following [3], we let  $D(f)$ ,  $R_0(f)$ , and  $R_2(f)$  respectively denote the deterministic, zero-error, and bounded-error classical query complexities of  $f$ , and let  $Q_E(f)$ ,  $Q_0(f)$ , and  $Q_2(f)$  denote the corresponding quantum query complexities. We have:

- $n \geq D(f) \geq Q_E(f) \geq Q_0(f) \geq Q_2(f)$  and  $n \geq D(f) \geq R_0(f) \geq R_2(f) \geq Q_2(f)$ ,
- $D(f) = O(R_2(f)^3)$  and  $D(f) = O(R_0(f)^2)$  [5],
- $D(f) = O(Q_2(f)^6)$ , or  $O(Q_2(f)^4)$  if  $f$  is monotone [3], and
- $D(f) = O(Q_0(f)^4)$  [4].

Let  $\mu$  be the uniform distribution over  $\{0,1\}^n$ . Following Ambainis and de Wolf [2], we let  $D^\mu(f)$  be the average-case deterministic query complexity under the uniform distribution of inputs. (Note that in [2],  $\mu$  can be non-uniform, whereas here it is always uniform.) The average-case bounded-error analogs of  $D^\mu(f)$ ,

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$R_2^\mu(f)$  and  $Q_2^\mu(f)$  in the classical and quantum settings respectively, can be super-exponentially smaller than  $D^\mu(f)$  [2]. On the other hand, we have  $D^\mu(f) = R_0^\mu(f)$  by Yao's minimax principle: viewing the questioner's choice of query algorithm and the oracle's choice of response algorithm in matrix-game terms, if the oracle is committed to a fixed randomized strategy (as it is in the average-case setting), then the questioner has nothing to gain by using randomization (assuming the questioner's goal is the same, namely to evaluate  $f$  with probability 1). Therefore we need not consider  $R_0^\mu(f)$ .

Here we show that  $D^\mu(f) = O(Q_2(f)^5)$ , or  $O(Q_2(f)^3)$  if  $f$  is monotone. The proof has two components. Theorem 1 gives a deterministic classical algorithm for evaluating  $f$  with few queries in the average case, yielding an upper bound on  $D^\mu(f)$ . The theorem is a refinement of [3, Lemma 5.3], which gives an upper bound on  $D(f)$ . Theorem 2 gives a lower bound on  $Q_2(f)$  in terms of the expected block sensitivity. The bound is obtained via the quantum adversary argument, which was recently introduced by Ambainis [1].

Given  $X \in \{0, 1\}^n$  and a block  $B$  of variables, let  $X(B)$  be the input obtained from  $X$  by flipping the values of all the variables in  $B$ . Following [5, 3]:

- A *1-certificate* is an assignment  $C : B \rightarrow \{0, 1\}$  of values to a block  $B$  of variables, such that  $f(X) = 1$  whenever  $X$  is consistent with  $C$ . The size of  $C$  is  $|B|$ . A *0-certificate* is defined similarly. A 1-certificate  $C$  is *minimal* if no proper sub-block  $C'$  of  $C$  is a 1-certificate (similarly for 0-certificates). The *certificate complexity*  $C_X(f)$  of  $X$  is the size of the smallest  $f(X)$ -certificate that agrees with  $X$ . The certificate complexity  $C(f)$  of  $f$  is the maximum of  $C_X(f)$  over all  $X$ .
- The *block sensitivity*  $\text{bs}_X(f)$  of  $X$  is the maximum number  $b$  of disjoint blocks  $B_1, \dots, B_b$  of variables such that for all  $1 \leq i \leq b$ ,  $f(X) \neq f(X(B_i))$ . The block sensitivity  $\text{bs}(f)$  of  $f$  is the maximum of  $\text{bs}_X(f)$  over all  $X$ .

Let  $C_\mu(f) = E_\mu[C_X(f)]$  be the mean of  $C_X(f)$  over all  $X$ . Likewise let  $\text{bs}_\mu(f) = E_\mu[\text{bs}_X(f)]$  be the mean of  $\text{bs}_X(f)$  over all  $X$ . Let  $\text{bs}_\mu^{(1)}(f) = E_{\mu(1)}[\text{bs}_X(f)]$  be the mean block sensitivity among  $X$  such that  $f(X) = 1$ , and let  $\text{bs}_\mu^{(0)}(f) = E_{\mu(0)}[\text{bs}_X(f)]$  be the mean block sensitivity among  $X$  such that  $f(X) = 0$ .

### 3 Results

First we relate  $D^\mu(f)$  to the mean block sensitivity  $\text{bs}_\mu(f)$ , along the lines of [3, Lemma 5.3].

**Theorem 1**  $D^\mu(f) \leq 2 \text{bs}_\mu(f) C(f)$ .

**Proof.** Let a *satisfying* certificate be one that agrees with  $X$ ; let a *consistent* certificate be one that agrees with the  $X$ -values queried so far. The following algorithm returns a satisfying 0-certificate in expected number of queries at most  $\text{bs}_\mu^{(0)}(f)C(f)$ , assuming that  $f(X) = 0$  (the expectation is over the uniform distribution of all  $X$  satisfying this condition).

Choose a minimal consistent 1-certificate and query those of its variables whose  $X$ -values are still unknown. Repeat until a satisfying 0-certificate is found among the variables that have been queried.

Call this algorithm  $A_0$ .  $A_0$  can be made deterministic by choosing certificates in some fixed lexicographic order. To see that  $A_0$  always returns a satisfying 0-certificate, note that, for the special case  $f(X) = 0$  that we're considering,  $A_0$  reduces to Algorithm  $A$  of [3, Lemma 5.3].  $A$  always returns a satisfying 0-certificate when  $f(X) = 0$ , therefore so does  $A_0$ .

It remains to show that the expected number of queries used by  $A_0$  is at most  $\text{bs}_\mu^{(0)}(f)C(f)$ . Suppose that, after  $A_0$  has queried  $k$  1-certificates,  $C_1, \dots, C_k$ , no satisfying 0-certificate has yet been found. Then

there exists a  $Y$  consistent with the bits queried so far such that  $f(Y) = 1$ . Furthermore,  $Y$  contains a satisfying 1-certificate  $C_{k+1}$ . We will derive from these  $C_i$  disjoint blocks  $B_i \subseteq X$  such that  $f$  is sensitive to each  $B_i$  on  $X$ . For each  $1 \leq i \leq k+1$ , let  $B_i$  be the set of variables on which  $X$  and  $C_i$  disagree. Clearly each  $B_i$  is non-empty. Now,  $X(B_i)$  agrees with  $C_i$ , therefore  $f(X(B_i)) = 1$ , so that  $f$  is sensitive to each  $B_i$  on  $X$ . Let  $v$  be a variable in some  $B_i$ ; then  $X^{(v)} = Y^{(v)} \neq C_i^{(v)}$ . For  $j > i$ ,  $C_j$  has been chosen consistent with all variables queried so far (including  $v$ ), so we cannot have  $X^{(v)} = Y^{(v)} \neq C_j^{(v)}$ , hence  $v \notin B_j$ . Therefore all  $B_i$  and  $B_j$  are disjoint. It follows that  $k$  (the number of 1-certificates queried) can be at most  $\text{bs}(X)$ .

Now, since the input is chosen uniformly at random among all  $X$  with  $f(X) = 0$ , the expectation of  $\text{bs}(X)$  is  $\text{bs}_\mu^{(0)}(f)$ . Therefore  $A_0$  returns a satisfying 0-certificate after querying an expected number of certificates at most  $\text{bs}_\mu^{(0)}(f)$ , or after an expected total number of queries at most  $\text{bs}_\mu^{(0)}(f)C(f)$ .

An analogous algorithm,  $A_1$ , returns a satisfying 1-certificate in expected number of queries at most  $\text{bs}_\mu^{(1)}(f)C(f)$ , assuming that a 1-certificate exists (i.e. that  $f(X) = 1$ ). Suppose that we interleave  $A_0$  and  $A_1$ , alternating between the two until either  $A_0$  or  $A_1$  halts and returns a certificate, and that when  $X$  is chosen from  $\mu$ ,  $f(X) = 1$  with probability  $p$ . Then the expected total number of queries is at most

$$2p \text{bs}_\mu^{(1)}(f)C(f) + 2(1-p) \text{bs}_\mu^{(0)}(f)C(f) = 2 \text{bs}_\mu(f)C(f). \quad \blacksquare$$

One can show, using a similar argument, that  $D^\mu(f) \leq 2 \text{bs}(f)C_\mu(f)$ . We conjecture that  $D^\mu(f) = O(\text{bs}_\mu(f)C_\mu(f))$ , but are unable to show this.

We next give a lower bound on  $Q_2(f)$  in terms of the mean block sensitivity. The proof is along the lines of Ambainis [1, Theorem 3]; for completeness, we recapitulate some of the material in that manuscript.

**Theorem 2**  $Q_2(f) \geq (1/2 - \sqrt{2}/3) \text{bs}_\mu(f)$ .

**Proof.** For each  $X$ , choose  $\text{bs}_X(f)$  disjoint minimal blocks  $B_1^{(X)}, \dots, B_{\text{bs}_X(f)}^{(X)}$  such that for all  $i$ ,  $f(X) \neq f(X(B_i^{(X)}))$ . (By minimal, we mean that for each  $i$ , no proper sub-block  $B'$  of  $B_i^{(X)}$  has the property that  $f(X) \neq f(X(B'))$ .) Call  $X(B_1^{(X)}), \dots, X(B_{\text{bs}_X(f)}^{(X)})$  the *block-neighbors* of  $X$ . (Note that if  $Y$  is a block-neighbor of  $X$ ,  $X$  is not necessarily a block-neighbor of  $Y$ .)

Let  $A$  be a quantum algorithm to evaluate  $f(X)$  with probability of error  $\varepsilon = 1/3$ . Following [1], instead of running  $A$  with a single string as input, we run  $A$  with the uniform superposition of all strings in  $\mu$  as input. Let  $\mathcal{H}_I$  be the ‘input subspace’ spanned by basis vectors  $|X\rangle$  corresponding to the possible inputs  $X$ . Let  $\rho_k$  be the density matrix of  $\mathcal{H}_I$  after  $A$  has made  $k$  queries. Let  $S_k$  be the sum of  $(\rho_k)_{X,Y}$  for all ordered pairs  $(X,Y)$  such that  $Y$  is a block-neighbor of  $X$ . Suppose that  $A$  makes a total number of queries  $T$ . Then:

1.  $S_0 = \text{bs}_\mu(f)$ . This is because there are  $2^n$  input strings, the mean number of block-neighbors of a string is  $\text{bs}_\mu(f)$ , and every entry of  $\rho_0$  is  $2^{-n}$ .
2.  $S_T \leq 2\sqrt{\varepsilon(1-\varepsilon)} \text{bs}_\mu(f) = (2\sqrt{2}/3) \text{bs}_\mu(f)$ , by [1, Lemma 1].
3.  $S_{k-1} - S_k \leq 2$ .

Together, these statements imply that  $T \geq (1/2 - \sqrt{2}/3) \text{bs}_\mu(f)$ .

We now prove the third statement. Express the state before the  $k^{\text{th}}$  query as

$$|\psi_{k-1}\rangle = \sum_{i,a,z,X} \alpha_{i,a,z,X} |i,a,z\rangle \otimes |X\rangle$$

where  $i$  is the index of the variable  $X_i$  being queried,  $a$  is a bit for recording the answer, and  $z$  is a collection of extra work bits. Then after the  $k^{th}$  query we have

$$|\psi_k\rangle = \sum_{i,a,z,X} \alpha_{i,a,z,X} |i, a \oplus X_i, z\rangle \otimes |X\rangle = \sum_{i,a,z,X} \alpha_{i,a \oplus X_i, z, X} |i, a, z\rangle \otimes |X\rangle.$$

Define

$$|\psi_{i,a,z}\rangle = \sum_X \alpha_{i,a,z,X} |X\rangle \text{ and } |\psi'_{i,a,z}\rangle = \sum_X \alpha_{i,a \oplus X_i, z, X} |X\rangle.$$

Then  $\rho_{k-1,i} = \sum_{a,z} |\psi_{i,a,z}\rangle \langle \psi_{i,a,z}|$  and  $\rho_{k,i} = \sum_{a,z} |\psi'_{i,a,z}\rangle \langle \psi'_{i,a,z}|$  are the components of  $\rho_{k-1}$  and  $\rho_k$  respectively corresponding to querying  $X_i$ . We can then represent  $\rho_{k-1}$  as  $\sum_{i=1}^n \rho_{k-1,i}$  and  $\rho_k$  as  $\sum_{i=1}^n \rho_{k,i}$ . Then  $S_{k-1} - S_k \leq \sum_{i=1}^n S_{k,i}$  where

$$S_{k,i} = \sum_{X,Y: Y \text{ a block-neighbor of } X} |(\rho_{k-1,i})_{X,Y} - (\rho_{k,i})_{X,Y}|.$$

The only entries that differ in  $\rho_{k,i}$  and  $\rho_{k-1,i}$  are the ones that correspond to  $X, Y$  with  $X_i \neq Y_i$ . For every  $X$ , there is at most one block-neighbor  $Y$  having this property. (The fact that we're dealing with block-neighbors, rather than with ordinary neighbors as in [1], doesn't change this.) Therefore

$$\sum_{X,Y: Y \text{ a block-neighbor of } X} (\rho_{k-1,i})_{X,Y} \leq \sum_{X,Y: Y \text{ a block-neighbor of } X} [(\rho_{k-1,i})_{X,X} + (\rho_{k-1,i})_{Y,Y}] / 2 \leq \sum_X (\rho_{k-1,i})_{X,X} = \text{Tr } \rho_{k-1,i}.$$

A similar result is true for  $\rho_{k,i}$ . So we have that  $S_{k,i} \leq \text{Tr } \rho_{k-1,i} + \text{Tr } \rho_{k,i}$  and that

$$S_{k-1} - S_k \leq \sum_i S_{k,i} \leq \sum_i (\text{Tr } \rho_{k-1,i} + \text{Tr } \rho_{k,i}) = 2. \quad \blacksquare$$

Combining Theorem 1 and Theorem 2 with  $C(f) \leq \text{bs}(f)^2$  [5] and  $\text{bs}(f) \leq 16Q_2(f)^2$  [3] we obtain

$$D^\mu(f) \leq 2 \text{bs}_\mu(f) C(f) \leq 12(3 + 2\sqrt{2}) \text{bs}(f)^2 Q_2(f) \leq 3072(3 + 2\sqrt{2}) Q_2(f)^5 \approx 17905 Q_2(f)^5.$$

When  $f$  is monotone,  $C(f) = \text{bs}_\mu(f)$  [5], so we obtain

$$D^\mu(f) \leq 2 \text{bs}_\mu(f) \text{bs}(f) \leq 192(3 + 2\sqrt{2}) Q_2(f)^3 \approx 1119 Q_2(f)^3.$$

## 4 Some Open Problems

- For the case of zero-error quantum algorithms, Buhrman et al. [4] showed that  $D(f) = O(Q_0(f)^4)$ . Can we relate  $D^\mu(f)$  to  $Q_0(f)$ ?
- What can we say when  $\mu$  is non-uniform?

## 5 Acknowledgments

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