**Chapter on Quantum for**
**Computational Intractability:**
**A Guide to Algorithmic Lower Bounds**
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**Note to Proofreaders** There will be passages like "In Chapter ?? we discussed..." This is not a mistake. This is a a result of giving you this chapter and not the rest of the book. We have not put it into the book yet.

# 1 Introduction

This book is about classical computing. The algorithms and reductions in this book can be carried out by a modern computer running on a single CPU or multiple CPUs. We will call such computers ***classical*** in that they are based on classical physics (at the bit level computers are based on electricity) and not quantum physics. We use the term ***classical algorithm*** for an algorithm that can be run on a classical computer. What we call a ***classical algorithm*** in this chapter was just an ***algorithm*** in all of the other chapters.

There are theoretical devices called ***quantum computers***. An algorithm that can be run on such a device is a ***quantum algorithm***. There are some problems for which, theoretically, there is a quantum algorithm that is faster than any (known) classical algorithm. Hence quantum algorithms are of interest (we later list other reasons they are of interest).

Quantum computing is a vast topic that we will, for reasons of space, only be able to discuss briefly. We will have very few definitions, proofs, algorithms, or reductions. For basic definitions and more information on quantum computing see (1) the references in this chapter, (2) many websites that you can get to by doing a web search on `Quantum Computing`, and (3) the books by Aaronson [1] or Nielson & Chung [44].

In this chapter we will state results about quantum algorithms and, in some cases, compare them to classical algorithms. We will also look at a quantum version of a classical problem. The topics and results chosen highlight when the classical world and the quantum world differ or seem to differ.

A cautionary note: from the popular press one might think that quantum computers, if they are built, can solve world hunger, predict the stock market, and solve **NP**-complete problems. Most experts agree that this is not the case in reality. This misunderstanding may come from a misinterpretation of the *many-worlds interpretation of quantum mechanics* which gives the false impression that quantum algorithms are massively parallel. They are not. In reality there seem to be only a few problems of interest that quantum computers (if they are built) can do much faster than classical computers. We also note that most experts in quantum do not think that **NP**-complete problems can be solved quickly by a quantum computer.

Given that the usefulness of quantum algorithms seems limited, why study them?

1. There are two problems, *Factoring* and *Discrete Log*, which (1) are very important, and (2) quantum algorithms for them are much faster (polynomial time versus exponential time) than any known classical algorithm.

2. There are many problems that have quantum algorithms that are faster than any known classical algorithm. Childs & Dam [19] survey algebraic problems that have quantum algorithms which seem faster than any known classical algorithm. Jordan [35] maintains a website of problems that have quantum algorithms that seem faster than any known classical algorithm.

The speedups are usually not that large. Even so, they are a proof-of-concept for quantum algorithms being useful.

3. Richard Feynman first conceived of quantum computing as a way to potentially simulate quantum mechanics. This is another problem where quantum computers may outperform classical ones.

4. There have been cases where a classical algorithm was inspired by research on quantum algorithms. We give an example. Kerenidis & Prakash [38] had a quantum algorithm for a recommendation system. Tang, while trying to show that no classical algorithm could do as well as the quantum algorithm, found a classical one that did [52]. We stress that this classical algorithm was found because of research in quantum algorithms. For other examples do a websearch for *Quantum Inspired Classical Algorithms*.

5. The study of quantum algorithms has lead to results in classical computing. See the survey of Drucker & de Wolf [23] for examples.

6. The attempt to build quantum computers may lead to interesting insights into quantum physics. See next point.

7. The most exciting development that could happen would be if the attempt to build quantum computers leads to a discovery that the current theories of quantum mechanics are wrong or incomplete.

# 2 Factoring

Factoring is an important problems for cryptography. Many cryptosystems would be broken if factoring is easy. Hence, in contrast to work in algorithms, cryptographers hope that factoring is hard.

We will define factoring slightly differently than how it was defined in Chapter **??**.

## 2.1 Classical Factoring

**Problem 2.1.** FACTORING *(*FACT*)*

*INSTANCE: A number $N$*

*QUESTION: If $N$ is prime then output* **PRIME**. *If $N$ is not prime then output a non-trivial factor of $N$.*

As noted in Chapter **??** there are no polynomial-time algorithms for FACT, nor is there a proof that its **NP**-complete. There are reasons to think it is not **NP**-complete (See Exercise **??**).

Algorithms for factoring are hard to analyze and depend on (widely believed) conjectures in number theory. The fastest known algorithm, the General Number Field Sieve, is believed to have running time roughly $2^{1.93\, L^{1/3}(\lg L)^{2/3}}$, where $L = \lg N$ is the length of the input number $N$. This bound is small enough that the algorithm is practical for moderately large inputs. The naive algorithm for factoring takes time $2^{L/2}$. Hence reducing the time to roughly $2^{L^{1/3}}$ is a real improvement. The first one to come up with the idea for the Number Field Sieve is Pollard, though his work was never published. See the survey of Pomerance [46] or the collection of articles on the Number field Sieve edited by A. Lenstra & H. Lenstra [42].

There is no clear consensus on whether FACT is in **P**: in Gasarch's [26] 2019 Poll on **P** vs **NP** he also asked about FACT. Of the 108 people who responded 38 (35%) thought FACT $\in$ **P**, while

70 (65%) though FACT $\notin \mathbf{P}$. It's been said that cryptographers think (hope?) that FACT $\notin \mathbf{P}$ while number theorist think FACT $\in \mathbf{P}$.

If FACT $\in \mathbf{P}$ this will require new techniques. Here is why:

1. The last improvement in factoring algorithms was the Number Field Sieve in 1988.
2. There are reasons to think that the current methods yield algorithms with running times of the form $2^{L^t(\ln L)^{1-t}}$, where $0 < t < 1$. The General Number Field Sieve achieves $t = 1/3$. It is plausible that current techniques will solve FACT in time (say) $2^{L^{1/10}(\ln L)^{9/10}}$ but not in $\mathbf{P}$.

For more about factoring, see Wagstaff's book [55].

### 2.2 Quantum Factoring

What about Quantum Polynomial time?

**Theorem 1.** *Let the input to a factoring algorithm be $N$. Let $L = \lg N$ which is the length of $N$.*

1. *(Shor [49, 50]) There is a polynomial quantum algorithm for* FACT.
2. *(Beckman et al. [9]) There is quantum algorithm for* FACT *that takes time $O(L^2 \log(L) \log(\log(L)))$. (This paper used Shor's algorithm as a starting point.)*

We note the following

1. The key quantum component of Shor's algorithm for FACT is the quantum Fourier transform.
2. The constants in Beckman et al.'s version of Shor's algorithm are small. The biggest obstacle to running the algorithm is building a quantum computer that can handle many qubits.
3. Martin-Lopez et al. [43] have factored 21 on a quantum computer using Shor's algorithm. This can be considered a proof-of-concept. Other bigger numbers have been reported to have been factored by *a quantum computer* but they really used a lot of classical computing to set the problem up and hence we do not count those. Smolin et al. [51] discuss this issue.

**Upshot** If FACT $\notin \mathbf{P}$ then FACT will be an example of a problem that quantum computers can do faster than classical computers. Proving FACT $\notin \mathbf{P}$ is hard since it implies $\mathbf{P} \neq \mathbf{NP}$.

## 3 Discrete Log

Discrete Log is an important problems for cryptography. Many cryptosystems would be broken if Discrete Log is easy. Hence, in contrast to work in algorithms, cryptographers hope that Discrete Log is hard.

We will define Discrete Log slightly differently than how it was defined in Chapter **??**.

### 3.1 Classical Discrete Log

**Problem 3.1.** DISCRETE LOG (DL)

*INSTANCE: A prime $p$, a generator $g$ of $\mathbb{Z}_p$, and $a \in \mathbb{Z}_p$. ($g$ is a generator if $\{g, g^2, \ldots, g^{p-1}\} = \{1, \ldots, p-1\}$. Note that both sets are unordered so we are not saying $g = 1$.)*

*QUESTION: Find $x$ such that $g^x \equiv a \pmod{p}$.*

As noted in Chapter **??** there are no polynomial-time algorithms for DL, nor is there a proof that its **NP**-complete. There are reasons to think it is not **NP**-complete (See the discussion of DL in Chapter **??**.)

Algorithms for DL are hard to analyze and depend on (widely believed) conjectures in number theory. The fastest known algorithm, the Function Field Sieve, is believed to have running time roughly $2^{1.53\,L^{1/3}(\lg L)^{2/3}}$, where $L = \lg N$ is the length of the input number $N$. This bound is small enough that the algorithm is practical for moderately large inputs. The naive algorithm for DL takes time $2^L$. Hence reducing the time to roughly $2^{L^{1/3}}$ is a real improvement. Adleman [4] developed the Function field sieve and then elaborated the ideas with Huang (see [5]).

There is no formal connection between DL and FACT; however, the techniques for one seem to apply to the other. Hence the following two points made about FACT are true of DL also: (1) there is no consensus about if DL $\in$ **P**, and (2) if DL $\in$ **P** then this will require new techniques.

### 3.2 Quantum Discrete Log

What about Quantum Polynomial time?

**Theorem 2.** *Let the input to a* DL *algorithm be $N$. Let $L = \lg N$ which is the length of $N$.*

1. *(Shor [49, 50]) There is a polynomial quantum algorithm for* DL.
2. *(Folklore though can be obtained from Beckman et al. [9].) There is quantum algorithm for* DL *that takes time $O(L^2 \log(L) \log(\log(L)))$.*

We note the following

1. The key quantum component of Shor's algorithm for DL is the quantum Fourier transform. In Section 3.1 we noted that while there is no formal connection between DL and FACT, improvements in one tend to lead to improvements in the other. We were referring to classical algorithms. However, the same seems to be true for quantum algorithms: both the quantum algorithm for FACT and for DL use the quantum Fourier transform.
2. There do not seem to be any attempts to execute Shor's DL algorithm on a quantum computer. Since the quantum algorithms for FACT and DL are similar, the same techniques that were used for FACT will work on DL. Hence it is likely that a quantum computer could be used to find DL when $p, g, a$ are all $\leq 21$.

**Upshot** If DL $\notin$ **P** then DL will be an example of a problem that quantum computers can do much faster than classical computers. Proving DL $\notin$ **P** is hard since it implies **P** $\neq$ **NP**.

## 4 The Search Problem

**Problem 4.1.** SEARCH

*INSTANCE: Access to a function $f\colon \{0,\ldots,N-1\} \to \{0,1\}$. We are promised that there is only one $x$ such that $f(x) = 1$. We think of this function as representing a 1-element subset of $\{0,\ldots,N-1\}$*

*QUESTION: Find the $x$ such that $f(x) = 1$.*

*NOTE: The basic unit of computation is an evaluation of $f$ which we call a query.*

**Theorem 3.** *Let $N \in \mathbb{N}$ and $f\colon \{0,\ldots,N-1\} \to \{0,1\}$.*

1. *(Easy) There is a classical algorithm for* SEARCH *that takes $N$ queries in the worst case and $\frac{N}{2}$ queries in the average case. No deterministic or randomized algorithm can do* SEARCH*faster.*

2. *(Grover [30]) There is a quantum algorithm for* SEARCH *that uses $O(\sqrt{N})$ queries.*

3. *(Bennett et al. [10]) Any quantum algorithm for* SEARCH *requires $\Omega(\sqrt{N})$ queries.*

4. *If instead of having only 1 $x$ with $f(x) = 1$ there are $M$, and the goal is to find one of them, then this has classical complexity $O(N - M)$ queries and quantum complexity $O(\sqrt{N/M})$.*

**Upshot** SEARCH, with the complexity measure number-of-queries, is a problem where quantum computers are **provably** faster than classical computers.

# 5 The Traversal Problem

**Problem 5.1.** TRAVERSAL

*INSTANCE: A graph $G$ with $\Theta(2^n)$ vertices represented by $\Theta(n)$-bit strings. There are two distinguished vertices* ENTRANCE *and* EXIT*. The label of* ENTRANCE *(e.g., Vertex 0110) is given. The label of* EXIT *is not given.*

*QUESTION: Find the label of* EXIT*.*

*NOTE: The graph is large so it is not the input as is. Instead the input is through an oracle: the algorithm can ask, given a string $w$ of $\Theta(n)$ bits, return the following information:*

- *Is $w$ a vertex?*
- *If $w$ is a vertex then output all of its neighbors.*
- *If one of the neighbors of $w$ is* EXIT *then indicate this.*

*The number of queries is the complexity of the algorithm.*

*NOTE:* TRAVERSAL*is only asking to find* EXIT*. It is not asking to find the path from* ENTRANCE *to* EXIT*. We will later comment on this perhaps harder problem of having to find the path.*

This problem looks like it requires $\Omega(2^n)$ queries for either classical or quantum algorithms. And indeed, for the case of general graphs, that is the case. But there is a sequence of graphs where there is a large difference between classical algorithms and quantum algorithms.

One technique that a classical algorithm can use is a CLASSICAL RANDOM WALK: the algorithm picks a random neighbor of ENTRANCE, then a random neighbor of that neighbor, etc, until it finds EXIT. There is also a notion of a QUANTUM RANDOM WALK which we will not define.

**Theorem 4.**

1. *(Childs et al. [21]) There is a sequence $\{G_n\}_{i=1}^{\infty}$ such that the following hold: (a) $G_n$ has $\Theta(2^n)$ vertices, (b) there is a* QUANTUM RANDOM WALK *algorithm that solves* TRAVERSAL *on $G_n$ in time $O(n)$, (c) any* CLASSICAL RANDOM WALK *algorithms requires $2^{\Omega(n)}$ time.*

2. *(Childs et al. [20]) There is a sequence $\{G_n\}_{i=1}^{\infty}$ such that the following hold: (a) $G_n$ has $\Theta(2^n)$ vertices, (b) there is a* QUANTUM RANDOM WALK *algorithm that solves* TRAVERSAL *on $G_n$ in time $O(n)$, (c) any classical algorithms (whether or not it uses* CLASSICAL RANDOM WALK*) requires $2^{\Omega(n)}$ time.*

3. *(Childs et al. [18]) (Informal) Consider the problem of actually finding the path from* ENTRANCE *to* EXIT. *Under reasonable assumptions, any* CLASSICAL RANDOM WALK *or* QUANTUM RANDOM WALK *algorithm requires exponential time.*

**Upshot** TRAVERSAL, with the complexity measure number-of-queries, is a problem where quantum computers are **provably** faster than classical computers.

# 6 Quantum Streaming Algorithms

## 6.1 Classical Streaming for Triangle Counting and Distinguishing

**Problem 6.1.** TRIANGLE COUNTING TC
*INSTANCE: Graph $G = (V, E)$*
*QUESTION: Approximate the number of triangles in $G$.*

A related problem that is usually considered in the literature is that of TRIANGLE DISTINGUISHING, which is defined as follows.

**Problem 6.2.** TRIANGLE DISTINGUISHING TD
*INSTANCE: Graph $G = (V, E)$, a number $T$, and the promise that $G$ has either 0 triangles or $T$ triangles.*
*QUESTION: Does $G$ have 0 triangles or $T$ triangles?*

Clearly TD $\leq$ TC. Hence, a lower bound on TD implies a lower bound on TC.

We will state lower bounds on TD (and hence TC). We now state the problems that are used for to obtain these lower bounds.

**Definition 1.** *Let $n \in \mathbb{N}$.*

1. *A **perfect matching $M$ over $[2n]$** is a set of $n$ ordered pairs $(i, j)$, where $i$ and $j$ are distinct elements of $[2n]$, such that every $\ell \in [2n]$ is in exactly 1 ordered pair.*

2. *Let $M$ be a perfect matching $M$ over $[2n]$. We identify $M$ with the following $n \times 2n$ matrix: For every ordered pair $(i, j)$ in the matching there is a row with 1's in the $i$th and $j$th spot, and 0's everywhere else. Note that a perfect matching can be associated to many different matrices. We will turn this around: we will give Bob a perfect matching by giving him a matrix.*

**Problem 6.3.** BOOLEAN HIDDEN MATCHING BHM
*INSTANCE: Alice gets a string $x \in \{0,1\}^{2n}$. Bob gets (a) a perfect matching $M$ over $[2n]$ via a matrix as described in Definition 1, and (b) a string $w \in \{0,1\}^n$ where $w$ is promised to satisfy either $Mx = w$ or $Mx = \overline{w}$ (where $\overline{w}$ is $w$ with every bit flipped).*
*QUESTION: Determine which is the case: $Mx = w$ or $Mx = \overline{w}$.*

**Theorem 5.** *(Gavinsky et al. [27]) The randomized 1-way communication complexity, with Alice sending, is $\Omega(\sqrt{n})$.*

**Notation 1.** *Let $n$ denote the number of vertices, $m$ denote the number of edges, and $T$ is as in the problem statement. $\Delta_V$ (respectively $\Delta_E$) is the maximum number of triangles in $G$ that share a vertex (respectively an edge).*

The following are known.

**Theorem 6.**

1. *(Jayaram & Kallaugher [33]) There is a single-pass streaming algorithm for* TC *that uses space* $\tilde{O}\left(\frac{m\Delta_E}{T} + \frac{m\sqrt{\Delta_V}}{T}\right)$.
2. *(Braverman et al. [13]) Any single-pass streaming algorithm for* TD *(and hence for* TC*) uses space* $\Omega\left(\frac{m\Delta_E}{T}\right)$. *This proof uses a reduction of* INDEX *to* TD.
3. *(Kallaugher and Price [37]) Any single-pass streaming algorithm for* TD *(and hence for* TC*) uses space* $\Omega\left(\frac{m\sqrt{\Delta_V}}{T}\right)$. *This proof uses a reduction of* BHM *to* TD.
4. *Any single-pass streaming algorithm for* TD *(and hence for* TC*) requires space* $\Omega\left(\frac{m\Delta_E}{T} + \frac{m\sqrt{\Delta_V}}{T}\right)$. *This follows from Parts 2 and 3. Note that we now have matching bounds for one-pass streaming algorithms for* TC.

### 6.2 Quantum Streaming for Triangle Counting and Distinguishing

Quantum streaming algorithms were first defined by Khadiev et al. [39] (see also Ablayev et al. [3]). We will discuss modifying the proofs of the lower bounds for streaming on TD and TC from Theorem 6 to obtain lower bounds for quantum streaming for these problems.

Theorem 6.2 used that INDEX has communication complexity $\Omega(n)$. Fortunately, Ambainis et al. [6] showed that INDEX also has quantum communication complexity $\Omega(n)$. Hence we have the following analog to Theorem 6.2 by the same proof:

**Theorem 7.** *Any single-pass quantum streaming algorithm for* TD *(and hence for* TC*) requires space* $\Omega\left(\frac{m\Delta_E}{T}\right)$. *This proof uses a reduction of* INDEX *to* TD. *This follows from Theorem 6.2 and the work of Ambainis et al. [6].*

Can we do the same for Theorem 6.3? No. Gavinsky et al. [27] showed that the quantum communication complexity of BHM is $O(\log n)$. Hence we do not have a non-trivial lower bound for TC or TD in the region where $\Delta_E = O(1)$ and $T = \Omega(n)$. Indeed, there is a quantum streaming algorithm that works well in that region. Kallaugher [36] showed the following.

**Theorem 8.** *Restrict* TC *to the graphs where* $\Delta_E = O(1)$, $\Delta_V = \Omega(T)$, *and* $T = \Omega(m)$. *There is a single-pass quantum streaming algorithm for* TC *that uses space* $\tilde{O}(m^{2/5})$.

**Open 1.** *Find a lower bound of the form* $\Omega(m^c)$ *for* TC *in the case where* $\Delta_E = O(1)$, $\Delta_V = \Omega(T)$, *and* $T = \Omega(m)$.

### 6.3 Classical Streaming for $k$-Clique Counting and Distinguishing

In this section, we define two problems for $k$-clique finding which are analogous to Triangle Counting and Triangle Distinguishing.

**Problem 6.4.** $k$-Clique Counting (kCC)
*INSTANCE: Graph* $G = (V, E)$ *and* $k \in \mathbb{N}$.
*QUESTION: Approximate the number of cliques of size* $k$ *in* $G$.

**Problem 6.5.** *$k$-Clique Distinguishing (*kCD*)*

*INSTANCE: Graph $G = (V, E)$, $C \in \mathbb{N}$, and the promise that $G$ has either 0 $k$-cliques or $\geq C$ $k$-cliques.*

*QUESTION: Determine if $G$ has 0 $k$-cliques or $\geq C$ $k$-cliques.*

Clearly kCD $\leq$ kCC. Hence a lower bound on kCD implies a lower bound on kCC.

Theorem 6.2 stated a $\Omega\left(\frac{m\Delta_E}{T}\right)$ space lower bound for single-pass streaming algorithms for Triangle Distinguishing. A similar proof gives the same lower bound for $k$-Clique Distinguishing (with $T$ being the number of $k$-cliques); however this gives a trivial lower bound on most graphs, since $\Delta_E$ is usually small. We want a stronger lower bound for more general graphs. Additionally, since the quantum streaming complexity of triangle counting in the parameter setting $\Delta_E = O(1)$ and $T = \Omega(m)$ is an open problem it might be instructive to look for lower bounds on $k$-clique counting for $k \geq 4$ in this parameter setting to understand if the difficulty of this problem is unique for triangle counting.

For the next exercise you need the following definition and theorem.

**Definition 2.** *Let $k, n \in \mathbb{N}$.*

1. *A **perfect hypermatching $M$ over $[kn]$** is a set of $n$ ordered $k$-tuples $(i_1, \ldots, i_k)$, where $i_1, \ldots, i_k$ are distinct elements of $[kn]$, such that every $\ell \in [kn]$ is in exactly 1 ordered $k$-tuple.*

2. *Let $M$ be a perfect hypermatching $M$ over $[kn]$. We identify $M$ with the following $n \times kn$ matrix: For every ordered $k$-tuple $(i_1, \ldots, i_k)$ in the hypermatching there is a row with 1's in the $i_1$th, $i_2$th, ..., $i_k$th spot, and 0's everywhere else. Note that a perfect hypermatching can be associated to many different matrices. We will turn this around: we will give Bob a perfect hypermatching by giving him a matrix.*

**Problem 6.6.** Boolean Hidden Hypermatching BHHM

*INSTANCE: Alice gets a string $x \in \{0,1\}^{kn}$. Bob gets (a) a perfect hypermatching $M$ over $[kn]$ via a matrix as described in Definition 2, and (b) a string $w \in \{0,1\}^n$ where $w$ is promised to satisfy either $Mx = w$ or $Mx = \overline{w}$ (where $\overline{w}$ is $w$ with every bit flipped).*

*QUESTION: Determine which is the case: $Mx = w$ or $Mx = \overline{w}$.*

**Theorem 9.**

1. *(Verbin & Yu [54]) The randomized one-way communication complexity for* BHHM*, with Alice sending, is $\Omega(n^{1-(1/k)})$.*

2. *(Shi et al. [48]) the quantum one-way communication complexity for* BHHM*, with Alice sending, is $\Omega(n^{1-(2/k)})$.*

**Exercise 1.**

1. *Prove that any classical single-pass streaming algorithm for* kCD *requires space $\Omega\left(m^{1-1/k}\right)$. (Hint: Use the lower bound on* BHHM *from Theorem 9.1).*

2. *Prove that any quantum single-pass streaming algorithm for* kCD *requires space $\Omega\left(m^{1-2/k}\right)$ (space is measured in qubits). (Hint: Use the lower bound on* BHHM *from Theorem 9.2).*

**Open 2.**

1. *We have looked at counting and detecting triangles and $k$-cliques. Look at the problems of counting and detecting other subgraphs such as $k$-cycles.*

2. *Obtain classical and quantum upper and lower bounds on $p$-pass streaming algorithms.*

# 7 MIP* = RE

Lets consider 3COL $\in$ **NP** as an interactive proof involving two people: an all powerful prover and a poly time verifier. The prover wants to convince the verifier that a given graph is 3-colorable.

## 7.1 Separations

So far we have failed to obtain large separations between classical and quantum streaming algorithms. Note that we have been looking at *natural* streaming problems. There are results for contrived problems.

**Theorem 10.**

1. *(Le Gall [25]) There exists a streaming problem which (a) any classical algorithm requires $\Omega(n^{1/3})$ space, (b) there is a quantum algorithm that uses $O(\log n)$ space.*
2. *(Gavinsky et al. [27]) There exists a streaming problem which (a) any classical algorithm requires $\Omega(n^{1/2})$ space, (b) there is a quantum algorithm that uses $O(\log n)$ space.*

*(There are reasons why the first result is not quite comparable to the second result.)*

**Open 3.** *Find natural streaming problems for which there is a large separation between classical and quantum algorithms.*

Upshot Lower bounds on classical or quantum streaming algorithms are obtained by lower bounds on classical or quantum communication complexity. Hence the difficulty in obtaining a separation for streaming algorithms is to find a separation for communication complexity problems. This has been done for some contrived streaming problems; however, we would like to have a separation for natural problems.

# 8 MIP* = RE

Lets consider 3COL $\in$ **NP** as an interactive proof involving two people: an all powerful prover and a poly time verifier. The prover wants to convince the verifier that a given graph is 3-colorable.

- The prover sends the verifier a string $y$ that he hopes will convince the verifier that $x \in A$. The obvious thing to send is a 3-coloring of $G$.
- The verifier then determines if $y$ really is a 3-coloring of $G$. If so then he accepts that $G$ is 3-colorable. If not then he now believes $G$ is not 3-colorable.

Note that (a) there is only one prover, (b) the conversation is only one direction (prover sends a string to verifier), (c) the verifier is deterministic polynomial time, and (d) the verifier is convinced $G \in$ 3COL iff $G \in$ 3COL.

Interactive prove systems still have only one prover, but the conversation can be in rounds, the verifier can flip coins, and the verifier has a small probability of being wrong. Multiprover interactive proof systems also allow many provers, who cannot talk to each other.

Multiprover interactive proof systems have been used to get some lower bounds on how well a problem can be approximated in poly time, and can be considered a precursor to PCP and the Unique Games Conjecture.

**Definition 3.**

1. *A set $A$ is in **MIP** if there is a Multiprover interactive system such that (a) if $x \in A$ then the verifier accepts, and (b) if $x \notin A$ then the verifier rejects with probability $\geq 0.9$.*

2. *If we allow the provers to share entangled quantum states (the verifier is still classical) then this is* **MIP***.*

MIP and MIP* differ a lot:

**Theorem 11.**

1. *(Babai et al. [8]) MIP = NEXP.*

2. *(Ji et al. [34]) MIP* = RE where RE is the first level of the arithmetic hierarchy, Since RE contains the Halting set, MIP* contains sets that are undecidable.*

**Upshot** MIP and MIP* give an example where in a classical setting problems are in NEXP and in quantum setting, problems can be undecidable.

# 9 Quantum Games

In the previous sections we measured how well an algorithm did by how much time or space it used (queries can be considered time). In this section we look at games and measure how well the players do by looking at their probability of winning.

We discuss two games such that if the players play the game with quantum resources they can provably do better than if they play the game with classical resources. For further discussion of these games, and other games with this property, see the survey of Brunner et al. [15].

## 9.1 The CHSH Game

Clauser, Horne, Shimony, and Holt [22] invented the CHSH Game as a realizable experiment that can differentiate quantum from classical computing. (They did not give it that name; however, named using their initials.) We note that Clauser won the 2022 Nobel prize in Physics for this and other work [47].

**Problem 9.1.** *The* CHSH Game
*INSTANCE: Alice gets bit $x$, Bob gets bit $y$. Before they get their bits they can discuss strategy.*
*QUESTION: Alice outputs bit $a$, Bob outputs bit $b$. Alice and Bob win iff $x \wedge y = a \oplus b$.*

Clauser et al. [22] proved the following (see also Aaronson [2, Chapter 13] for an exposition).

**Theorem 12.**

1. *If Alice and Bob play the* CHSH Game *with classical resources (a) there is a deterministic strategy where they win with probability* 0.75 *(both always output 0), (b) there is no strategy, deterministic or randomized, that does better.*

2. *If Alice and Bob play the* CHSH Game *with quantum resources (they prepare entangled qubits before the game begins) then (a) there is a strategy where they win with probability* $0.5+\sqrt{2}/4 \sim 0.85$ *(this is complicated), (b) there is no strategy that does better.*

This game is of interest since it is a case where the quantum world is provably different from the classical world. Note that the gap between the classical and quantum is $0.85 - .0.75 = 0.1$.

## 9.2 Magic Square Game

Cabello [16] defined the Magic Square Game, though he did not call it that. For more information on it also see the survey of Brassard et al. on Quantum pseudo-telepathy [12, Section 5] or the Wikipedia entry on Quantum pseudo-telepathy [56].

**Problem 9.2.** *The* Magic Square Game *(*MS Game*)*

*INSTANCE: Alice gets $i \in \{1,2,3\}$, Bob gets $j \in \{1,2,3\}$. They interpret $i$ as the row of a $3 \times 3$ matrix and $j$ as the column of a $3 \times 3$ matrix. Alice and Bob get to discuss strategy ahead of time.*

*QUESTION: Alice and Bob both output a three-bit sequence. Alice's sequence is used as the $i$th row of a matrix. Bob's sequence is used as the $j$th column of a matrix. If the following three conditions hold then Alice and Bob win, else they lose. (a) The values in row $i$ add to an even number, (b) The values in column $j$ add to an odd number. (c) Alice and Bob's values are consistent (they agree at $(i,j)$).*

**Theorem 13.**

1. *If Alice and Bob play the* MS Game *with classical resources (a) there is a deterministic strategy where they win with probability $\frac{8}{9} = 0.88\ldots$, (b) there is no strategy, deterministic or randomized, that does better.*

2. *If Alice and Bob play the* MS Game *with quantum resources (they prepare entangled qubits before the game begins) then there is a strategy that wins with probably 1 (so always wins).*

This game is of interest since it is a case where the quantum world is provably different from the classical world. Note that the gap between the classical and quantum is $1.0 - 0.88\ldots = 0.11\ldots$.

Is there an interesting version of the MS Game on $k \times k$ matrices for $k \geq 4$? The following exercise shows that the answer is no.

**Exercise 2.**

1. *Give a $4 \times 4$ matrix $M$ of 0's and 1's such that every row sums to an even number and every column sums to an odd number.*

2. *Show that there is a classical strategy for the $4 \times 4$* MS Game *that wins with probability 1. (Hint: Use Part 1.)*

3. *Show that, for all $k \geq 4$, there is a classical strategy for the $k \times k$* MS Game *that wins with probability 1.*

## 9.3 Comparing The CHSH Game with The MS Game

We give two reasons why the MS Game game is better for distinguishing classical and quantum computation, and one reason why the CHSH Game game is better.

**Two reasons why the MS Game is better**

1. In the CHSH Game the gap between the classical and quantum players is 0.1. In the MS Game the gap between the classical and quantum players is 0.11 which is bigger!

2. For the MS Game the quantum players *always* win. This is better for repeated experiments. Assume the game is played $n$ times.

(a) For the MS GAME:
If Alice and Bob are classical then the expected number of wins is $8n/9$.
If Alice and Bob are quantum then the expected number of wins is $n$.
So if Alice and Bob lose just once, then they must be classical.

(b) For the CHSH GAME
If Alice and Bob are classical then the expected number of wins is $0.75n$.
If Alice and Bob are classical then the expected number of wins is $0.85n$.
These two cases are harder to distinguish since a lose by Alice and Bob could happen in either case.

**One reason why the CHSH Game is better.** Quantum computers (in 2023) are noisy. The computations are not that reliable, though there is a lot of work on quantum error correction to try to alleviate this. The CHSH GAME game is simpler and uses fewer operation, hence less noise. This is not just theoretical. The CHSH GAME is currently used in quantum systems in order to calibrate the quantum computers. The calculations done above for the MS GAME were assuming an error-free quantum computer which is not a reality yet.

In 2023 CHSH GAME is better and, as noted above, is actually used. However, there may be a time in the future where MS GAME is better.

**Upshot** There are games where if the players can use quantum entanglement then their probability of winning is provably higher than if they cannot.

# 10 Quantum MAXCUT

The title of this section might confuse people into thinking that there is a quantum algorithm for MAXCUT which is **NP**-complete. This is not the case. Instead, in this section we look at a quantum version of MAXCUT. This section differs from the previous ones in that rather than take a classic problem and see how well it can be solved with a quantum algorithm, we are taking a quantum problem, QMAXCUT, and seeing how well it can be solved with quantum techniques. We will first recall MAXCUT.

## 10.1 Classical MAXCUT

We state a variant of the classical MAXCUT problem. We will later see that the known upper and lower bounds for the MAXCUT problem we stated in Chapter **??** holds for this version.

**Definition 4.** *Let $G = (V, E, w)$ be a weighted graph. where all the weights are $\geq 0$ and sum to 1. We will view the weights as a probability distribution on the edges. A **cut** is a function, $f : V \to \{1, -1\}$. The **value** of a cut is given by*

$$\mathbb{E}\left[\frac{1}{2} - \frac{1}{2}f(u)f(v)\right] \tag{1}$$

*where the expected value is over the edges of $G$ via the distribution. Note that a cut can be viewed as assigning to each vertex a bit, even though the bits are in $\{-1, 1\}$ rather than the traditional $\{0, 1\}$.*

**Problem 10.1.** MAXCUT
*INSTANCE: $G = (V, E, w)$ a weighted graph where all the weights are $\geq 0$ and sum to 1.*
*QUESTION: Find the value of the largest cut.*

**Exercise 3.** *Show that the problem* MAXCUT *defined in this section is equivalent to the* MAXCUT *problem defined in Chapter* **??**.

By Exercise 3 all of the known lower bounds stated for MAXCUT earlier in this book hold here. This yields the first two items in the next theorem.

**Theorem 14.**

1. *(Hastad [31, Theorem 8.2] and Trevisan et al. [53, Theorem 4.4]) Assume* $\mathbf{P} \neq \mathbf{NP}$. *There does not exist an* $\epsilon > 0$, *and a polynomial time algorithm, that returns* $\geq (\frac{16}{17} + \epsilon)\text{OPT}$. *Note that* $\frac{16}{17} \sim 0.9411$.

2. *(Khot et al. [40], O'Donnell & Wu [45], Khot & O'Donnell [41]) Assume the Unique Games Conjecture holds. There does not exist an* $\epsilon > 0$, *and a polynomial time algorithm, that returns* $\geq (0.87856\ldots + \epsilon)\text{OPT}$. *See Chapter* **??** *for the exact constant.*

3. *(Goemans & Williamson [29]) There is an algorithm that matches the lower bound in Part 2. The algorithm given in Chapter* **??** *for* MAXCUT *can easily be modified for the version given in this section. Recall that the algorithm used a Semi Definite Program (*SDP*).*

In summary, the SDP approach to (classical) MAXCUT is optimal assuming the Unique Games Conjecture.

## 10.2 Quantum MAXCUT (qMAXCUT)

There is a quantum version of MAXCUT which we call QMAXCUT. We will define it and contrast it to the classical MAXCUT. For more background see either Carolan & Dontha [17] or the references in Theorem 15.

**Definition 5.** *A* ***unentangled qubit*** *is a size-two vector of complex numbers of unit 2-norm. (Intuitively, qubits are unentangled if they do not affect each other.)*

**Definition 6.** *Let* $G = (V, E, w)$ *be a weighted graph. where all the weights are* $\geq 0$ *and sum to 1. We will view the weights as a probability distribution on the edges. We give two definitions of* ***quantum cut****. The first one is a special case of the second one. The second one is the real one.*

1. *The product state case. There is no entanglement. In this case a* ***quantum cut*** *is an assignment of a qubit to every vertex. These qubits are not entangled. The* ***value*** *of a cut is given by*

$$\frac{1}{4}I - \frac{1}{4}\mathbb{E}\left[X_u X_v + Y_u Y_v + Z_u Z_v\right] \tag{2}$$

   *where the expected value is over the edges of* $G$, $I$ *is the identity matrices, and* $X, Y, Z$ *are Pauli matrices (which we will not define) acting on the* $u$*th qubit.*

2. *The general case. One can still assign to each vertex a qubit, but there is now potentially entanglement between qubits (we will not define entanglement). We must therefore interpret a cut as a quantum state on all* $|V|$ *qubits. The value of a general quantum cut is the same as in the product state case.*

**Problem 10.2.** QMAXCUT

*INSTANCE:* $G = (V, E, w)$ *a weighted graph where all the weights are* $\geq 0$ *and sum to 1.*
*QUESTION: Find the value of the largest quantum cut.*

**Theorem 15.** *The problem we are considering is* QMAXCUT.

1. *(Briët et al. [14], last page where they state* $u(3)$*) There is a polynomial time algorithm, that returns* $\geq (0.956\ldots)$OPT *for the restricted version where we only seek Product State Solutions. This algorithm uses* SDP *techniques. (See the paper for the exact constant.)*

2. *(Hwang et al. [32]) Assume the Unique Games Conjecture holds. The result in Part 1 is optimal for the restricted version where we only seek Product State Solutions.*

3. *(Gharibian & Parekh [28]) There is a polynomial time algorithm, that returns* $\geq (0.498\ldots)$OPT. *This algorithm uses* SDP *techniques.*

4. *(Hwang et al. [32]) If you use standard techniques for* SDP *based algorithm then, assuming the Unique Games Conjecture and a plausible inequality (a generalization of Borell's inequality to vectors—see Hwang [32, Conjecture 1.1]), you can do no better than the algorithm in Part 3.*

5. *(Anshu et al. [7]) There is a polynomial time algorithm, that returns* $\geq (0.53\ldots)$OPT.

Parts 4 and 5 are interesting because they give the following contrast:

1. For approximating MAXCUT in polynomial time, SDP techniques are optimal (assuming the Unique Games Conjecture).

2. For approximating QMAXCUT in quantum polynomial time, SDP techniques are not optimal.

We give a potential contrast:

1. Feige et al. [24] showed that if MAXCUT is restricted to graphs of bounded degree then there are approximations better than that in 0.87856OPT, which beats the lower bound for the general case stated in Theorem 14. So MAXCUT is easier if the graphs are of low degree.

2. Brandao and Harrow [11] proved lower bounds on approximating QMAXCUT for graph of low degree. So QMAXCUT seems harder if the graphs are of low degree.

We close this section with the obvious open problem:

**Open 4.** *Obtain closer upper and lower bounds for how well* QMAXCUT *can be approximated in polynomial time.*

**Upshot** The problems MAXCUT and QMAXCUT differ in several ways: (a) techniques needed for approximating in poly time (assuming the Unique Games Conjecture), and (b) MAXCUT seems easier on low degree graphs, whereas QMAXCUT seems harder. These observations indicate differences between the classical world and the quantum world.

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