Two-message quantum interactive proofs are in PSPACE

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Abstract

We prove that QIP(2), the class of problems having two-message quantum interactive proof systems, is a subset of PSPACE. This relationship is obtained by means of an efficient parallel algorithm, based on the multiplicative weights update method, for approximately solving a certain class of semidefinite programs.

1 Introduction

Since their introduction roughly 25 years ago [Bab85, GMR85], interactive proof systems have become a fundamental notion in the theory of computational complexity. The expressive power of one of the most basic variant of the interactive proof system model, wherein a polynomial-time probabilistic verifier interacts with a computationally unbounded prover for a polynomial number of rounds, is characterized [LFKN92, Sha92] by the well-known relationship

$$\text{IP} = \text{PSPACE}.$$  

Many variants of interactive proof systems have been studied, including public-coin interactive proof systems (or Arthur–Merlin games) [Bab85, BM88, GS89], zero-knowledge interactive proofs [GMR89, GMW91] and multi-prover interactive proofs [BOGKW88].

This paper is concerned primarily with quantum interactive proof systems, which are defined in a similar way to ordinary interactive proof systems except that the prover and verifier may perform quantum computations. Like their classical analogues, several variants of quantum interactive proof systems have been studied, including ordinary quantum interactive proofs [Wat03, KW00], public-coin quantum interactive proofs [MW05], zero-knowledge quantum interactive proofs [Wat02, Wat06, Kob08, HKSZ08], and multi-prover quantum interactive proofs [KM03, KKMV08]. The complexity class QIP of problems having quantum interactive proof systems is known [KW00] to satisfy

$$\text{PSPACE} \subseteq \text{QIP} \subseteq \text{EXP}.$$  

The containment QIP \(\subseteq\) EXP follows from the existence of polynomial-time algorithms for approximately solving semidefinite programs [GLS93]. (Somewhat simpler proofs of the containment QIP \(\subseteq\) EXP follow from the results of [BAT09, Wat09b], but these proofs still require efficient algorithms for solving convex-semidefinite programs.)

Quantum interactive proof systems have an interesting property that classical interactive proof systems are conjectured not to hold, which is that they can be parallelized to a constant number

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of rounds of interaction [KW00]. More precisely, it holds that QIP(3) = QIP, where in general QIP(m) denotes the class of problems having quantum interactive proof systems in which m messages are exchanged between the prover and verifier (with the prover always sending the last message). This leaves four basic classes that are defined naturally by quantum interactive proof systems: QIP(0) = BQP, QIP(1) = QMA, QIP(2), and QIP(3) = QIP. Of these classes, QIP(2) seems to be the most mysterious. It is known that ⊕MIP* ⊆ QIP(2) [Weh06] and QSZK ⊆ QIP(2) [Wat02, Wat06]. Here, ⊕MIP* denotes the class of problems having one-round two-prover classical interactive proof systems in which the provers share quantum entanglement, answer one bit each, and the verifier accepts or rejects based on the parity of these bits; and QSZK denotes the class of problems having statistical zero-knowledge quantum interactive proof systems. No upper bound other than the trivial containment QIP(2) ⊆ QIP, which implies QIP(2) ⊆ EXP, was previously known.

In this paper we prove the new containment:

\[ \text{QIP}(2) \subseteq \text{PSPACE}. \]

Similar to QIP ⊆ EXP, this containment is proved using semidefinite programming; but this time the containment is achieved by using an NC algorithm rather than a sequential polynomial-time algorithm. Our algorithm is based on the multiplicative weights update method, which was developed by several researchers and is described in the survey [AHK05] and in the PhD thesis of Kale [Kal07]. In particular, our algorithm is based on a general method that was independently discovered by Arora and Kale [AK07] and Warmuth and Kuzmin [WK06]. The key aspect of this approach that makes it useful for proving QIP(2) ⊆ PSPACE is its parallelizability: it is an iterative method in which each iteration is easily parallelized, and is such that only a very small number of iterations is needed for an approximation that is accurate enough for our needs. A related approach was used by two of us [JW09] to prove the containment of a different quantum complexity class (called QRG(1)) in PSPACE, but the specific technical details of the simulations are rather different.

The rest of this paper has the following structure. We begin with Section 2 which includes a brief discussion of background information needed for the rest of the paper, including linear algebra notation and terminology. Section 3 introduces two-message quantum interactive proof systems and establishes a simple fact concerning their robustness with respect to error bounds. In Section 4 we present a semidefinite programming formulation of the maximum probability with which a verifier in a two-message quantum interactive proof system can be made to accept, and the actual simulation of QIP(2) in PSPACE is split into the three sections that follow: Section 5 presents an overview of the simulation, while Sections 6 and 7 describe in more detail its two most technical parts. The precision requirements of the entire simulation are discussed in Section 8 and the paper concludes with Section 9.

2 Preliminaries

2.1 Linear algebra notation and terminology

For complex vector spaces of the form \( \mathcal{X} = \mathbb{C}^N \) and \( \mathcal{Y} = \mathbb{C}^M \), we write \( \text{L}(\mathcal{X}, \mathcal{Y}) \) to denote the space of linear operators mapping \( \mathcal{X} \) to \( \mathcal{Y} \), which is identified with the set of \( M \times N \) complex matrices in the usual way. An inner product on \( \text{L}(\mathcal{X}, \mathcal{Y}) \) is defined as \( \langle A, B \rangle = \text{Tr}(A^*B) \) for all \( A, B \in \text{L}(\mathcal{X}, \mathcal{Y}) \), where \( A^* \) denotes the adjoint (or conjugate transpose) of \( A \). The notation \( \text{L}(\mathcal{X}) \) is shorthand for \( \text{L}(\mathcal{X}, \mathcal{X}) \), and the identity operator on \( \mathcal{X} \) is denoted \( 1_{\mathcal{X}} \) (or just \( 1 \) when \( \mathcal{X} \) is understood).
The following special types of operators are relevant to the paper:

1. An operator $A \in L(\mathcal{X})$ is Hermitian if $A = A^*$. We write $\lambda(A) = (\lambda_1(A), \ldots, \lambda_N(A))$ to denote the vector of eigenvalues of $A$, sorted from largest to smallest: $\lambda_1(A) \geq \lambda_2(A) \geq \cdots \geq \lambda_N(A)$.

2. An operator $P \in L(\mathcal{X})$ is positive semidefinite if it is Hermitian and all of its eigenvalues are nonnegative. The set of such operators is denoted $\text{Pos}(\mathcal{X})$. The notation $P \succeq 0$ also indicates that $P$ is positive semidefinite, and more generally the notations $A \preceq B$ and $B \succeq A$ indicate that $B - A \succeq 0$ for Hermitian operators $A$ and $B$.

3. A positive semidefinite operator $\Pi \in \text{Pos}(\mathcal{X})$ is a projection if all of its eigenvalues are either 0 or 1. (Sometimes such an operator is called an orthogonal projection, but we have no need to discuss the more general sort of projection.)

4. An operator $\rho \in L(\mathcal{X})$ is a density operator if it is both positive semidefinite and has trace equal to 1. The set of such operators is denoted $\text{D}(\mathcal{X})$.

5. An operator $U \in L(\mathcal{X})$ is unitary if $U^*U = I_\mathcal{X}$. The set of such operators is denoted $\text{U}(\mathcal{X})$.

Three operator norms are discussed in this paper: the trace norm, Frobenius norm, and spectral norm, defined as

$$
\|A\|_1 = \text{Tr} \sqrt{A^*A}, \quad \|A\|_2 = \sqrt{\langle A, A \rangle}, \quad \text{and} \quad \|A\| = \max\{\|Au\| : u \in \mathcal{X}, \|u\| = 1\}
$$

respectively, for each $A \in L(\mathcal{X})$. Alternately, these norms are given by the $1, 2$ and $\infty$ norms of the vector of singular values of $A$. For every operator $A$ it holds that $\|A\| \leq \|A\|_2 \leq \|A\|_1$. We also use the inequalities $|\langle A, B \rangle| \leq \|A\| \|B\|_1, \|AB\| \leq \|A\| \|B\|$, and $\|AB\|_1 \leq \|A\| \|B\|_1$ a few times in the paper. The fidelity function is defined as

$$
F(P, Q) = \|\sqrt{P}\sqrt{Q}\|_1
$$

for positive semidefinite operators $P$ and $Q$ of equal dimension.

A super-operator is a linear mapping of the form $\Phi : L(\mathcal{X}) \rightarrow L(\mathcal{Y})$, for spaces of the form $\mathcal{X} = \mathbb{C}^n$ and $\mathcal{Y} = \mathbb{C}^m$. The identity super-operator on $L(\mathcal{X})$ is denoted $I_{L(\mathcal{X})}$. The adjoint super-operator to $\Phi$ is the unique super-operator $\Phi^* : L(\mathcal{Y}) \rightarrow L(\mathcal{X})$ for which $\langle Y, \Phi(X) \rangle = \langle \Phi^*(Y), X \rangle$ for all $X \in L(\mathcal{X})$ and $Y \in L(\mathcal{Y})$.

The following special types of super-operators are relevant to the paper.

1. $\Phi : L(\mathcal{X}) \rightarrow L(\mathcal{Y})$ is completely positive if it holds that $(\Phi \otimes I_{L(\mathcal{W})})(P) \in \text{Pos}(\mathcal{Y} \otimes \mathcal{W})$ for every choice of $\mathcal{W} = \mathbb{C}^k$ and $P \in \text{Pos}(\mathcal{X} \otimes \mathcal{W})$.

2. $\Phi : L(\mathcal{X}) \rightarrow L(\mathcal{Y})$ is trace-preserving if $\text{Tr}(\Phi(X)) = \text{Tr}(X)$ for every $X \in L(\mathcal{X})$.

3. $\Phi : L(\mathcal{X}) \rightarrow L(\mathcal{Y})$ is a quantum operation (also called an admissible super-operator or a quantum channel) if it is both completely positive and trace-preserving.

2.2 Remarks on NC and parallel matrix computations

To prove that QIP(2) is contained in PSPACE, we will make use of various facts concerning parallel computation. First, let us recall the definition of two complexity classes based on bounded-depth circuit families: NC and NC(poly). The class NC contains all functions that can be computed by logarithmic-space uniform Boolean circuits of polylogarithmic depth, while the class NC(poly) contains all functions that can be computed by polynomial-space uniform families of Boolean
circuits having polynomial-depth. By restricting these classes to predicates we obtain classes of languages (or more generally promise problems).

There are a few facts about these classes that we will need. The first fact, which follows from [Bor77], is that for languages (or promise problems) we have \( \text{NC}(\text{poly}) \subseteq \text{PSPACE} \). (In fact it holds that \( \text{NC}(\text{poly}) = \text{PSPACE} \), but we only need a containment in one direction.) The second fact is that functions in these classes compose nicely. In particular, if \( F : \{0,1\}^* \rightarrow \{0,1\}^* \) is a function in \( \text{NC}(\text{poly}) \) and \( G : \{0,1\}^* \rightarrow \{0,1\}^* \) is a function in \( \text{NC} \), then the composition \( G \circ F \) is also in \( \text{NC}(\text{poly}) \). This follows from the most obvious way of composing the families of circuits that compute \( F \) and \( G \), along with the observation that \( |F(x)| \) can be at most exponential in \( |x| \).

Finally, we will make use of the fact that many computations involving matrices can be performed by \( \text{NC} \) algorithms. We will restrict our attention to matrix computations on matrices whose entries have rational real and imaginary parts. Numbers of this form, \( \alpha = \left( \frac{a}{b} + \frac{ic}{d} \right) \) for integers \( a, b, c, \) and \( d \), are sometimes referred to as Gaussian rationals. We assume any number of this form is encoded as a 4-tuple \( (a, b, c, d) \) using binary notation, so that the length of \( \alpha \) is understood to be the total number of bits needed for such an encoding.

It is known that elementary matrix operations, such as additions, multiplications, and inversions, can be performed in \( \text{NC} \). (The survey [Gat93], for instance, describes \( \text{NC} \) algorithms for these tasks.) We will also make use of the fact that matrix exponentials and spectral decompositions can be approximated to high precision in \( \text{NC} \). In more precise terms, we have that the following problems are in \( \text{NC} \):

**Matrix exponentials**

**Input:** An \( n \times n \) matrix \( M \), a positive rational number \( \varepsilon \), and an integer \( k \) expressed in unary notation (i.e., \( 1^k \)), such that \( ||M|| \leq k \).

**Output:** An \( n \times n \) matrix \( X \) such that \( ||\exp(M) - X|| < \varepsilon \).

**Spectral decompositions**

**Input:** An \( n \times n \) Hermitian matrix \( H \) and a positive rational number \( \varepsilon \).

**Output:** An \( n \times n \) unitary matrix \( U \) and an \( n \times n \) real diagonal matrix \( \Lambda \) such that

\[ ||M - U\Lambda U^*|| < \varepsilon. \]

**Singular-value decompositions**

**Input:** An \( n \times m \) matrix \( M \) and a positive rational number \( \varepsilon \).

**Output:** An \( n \times r \) matrix \( U \) with orthonormal columns, an \( m \times r \) matrix \( V \) with orthonormal columns, and an \( r \times r \) diagonal matrix \( \Sigma \) with positive diagonal entries such that

\[ ||M - U\Sigma V^*|| < \varepsilon. \]

Note that in these problems, the description of \( \varepsilon \) has roughly \( \log(1/\varepsilon) \) bits, which means that highly accurate approximations are possible in \( \text{NC} \). The fact that matrix exponentials can be approximated in \( \text{NC} \) as claimed follows by truncating the series

\[ \exp(M) = 1 + M + M^2/2 + M^3/6 + \cdots \]


to a number of terms polynomial in \( k \) and \( \log(1/\varepsilon) \). (This is not a very practical way to compute matrix exponentials, but it establishes the fact we need.) The fact that spectral and singular value decompositions can be approximated in NC follows from a composition of known facts: in NC one can compute characteristic polynomials and null spaces of matrices, perform orthogonalizations of vectors, and approximate roots of integer polynomials to high precision \([\text{Csa76, BGH82, BCP83, BOPKT86, Gat93, Nef94}]\).

\section{Two-message quantum interactive proof systems}

The purpose of this section is to introduce the class QIP(2), including its definition and a simple proof that it is robust with respect to error bounds. For a general discussion of quantum interactive proof systems, as opposed to the somewhat simplified case in which only two messages are exchanged, the reader is referred to \([\text{KW00, Wat09a}]\).

\subsection{Definition of two-message quantum interactive proofs}

To define the class QIP(2), we begin by defining a \textit{two-message quantum verifier} \( V \) as a classical polynomial-time algorithm that, on each input string \( x \), outputs the description of two quantum circuits: \( U_x \) and \( V_x \). The circuit \( U_x \) describes the verifier’s initial preparation of a state, part of which is sent to the prover, while the circuit \( V_x \) describes the verifier’s actions upon receiving a response from the prover. For the sake of simplicity, and without loss of generality, we assume that for every input string \( x \), the circuits \( U_x \) and \( V_x \) are both composed of gates from some finite, universal set of unitary quantum gates whose entries have rational real and imaginary parts. The number of qubits on which the circuits \( U_x \) and \( V_x \) act is assumed to be equal to \( 2p(n) \), where \( n = |x| \) and \( p \) is some polynomial-bounded function. The first \( p(n) \) qubits represent the communication channel between the prover and verifier, while the remaining \( p(n) \) qubits serve as the private memory of the verifier. (It is not really necessary that the number of message qubits and private memory qubits agree, but it causes no change in the computational power of the model.)

A two-message quantum prover \( P \) is simply a collection of quantum operations (or, equivalently, completely positive and trace preserving super-operators) \( \{\Psi_x : x \in \{0, 1\}^*\} \). Such a prover is compatible with a given verifier \( V \) if each operation \( \Psi_x \) acts on \( p(n) \) qubits for the function \( p \) mentioned above.

An interaction between a two-message verifier \( V \) and a compatible prover \( P \) on an input \( x \) proceeds as follows:

1. \( 2p(n) \) qubits are initialized in the \( |0\rangle \) state.
2. The circuit \( U_x \) is applied to all of the qubits.
3. The prover’s operation \( \Psi_x \) is applied to the first \( p(n) \) qubits.
4. The circuit \( V_x \) is applied to all of the qubits.
5. The first qubit is measured with respect to the standard basis, with the outcome 1 indicating \textit{acceptance} and 0 indicating \textit{rejection}.

Figure\[\] illustrates such an interaction.

Now, a promise problem \( A = (A_{\text{yes}}, A_{\text{no}}) \) is in QIP(2) if and only if there exists a two-message verifier \( V \) with the following completeness and soundness properties:

1. (Completeness) If \( x \in A_{\text{yes}} \), then there exists a prover \( P \) that causes \( V \) to accept \( x \) with probability at least \( 2/3 \).
2. (Soundness) If \( x \in A_{\text{no}} \), then every prover \( P \) that is compatible with \( V \) causes \( V \) to accept \( x \) with probability at most \( 1/3 \).

3. Robustness of QIP(2) with respect to error bounds

It was proved in [KW00] that quantum interactive proof systems with negligible completeness error are amenable to parallel repetition. This allows for an exponential reduction in error for quantum interactive proof systems with three or more messages, because such proof systems can be transformed to have perfect completeness by a different method. However, this method does not work for two-message quantum interactive proof systems, because the perfect-completeness transformation requires the addition of messages. So, we will require a different method of error reduction.

Assume that \( A \) is a promise problem in QIP(2) and that \( (V, P) \) is a two-message quantum interactive proof system for \( A \) with completeness and soundness probabilities bounded by \( a \) and \( b \), where \( a - b \geq 1/q \) for some polynomial-bounded function \( q \). We will define a new verifier \( V' \) that has completeness probability at least \( 1 - 2^{-r} \) and soundness probability at most \( 2^{-r} \), for any choice of a polynomial-bounded function \( r \). A description of \( V' \) follows.

1. Let \( s = 2rq \) and let \( t = 8rq^2s \). Run \( st \) independent, parallel executions of the protocol for \( V \), one for each pair \((i, j)\) with \( i \in \{1, \ldots, s\} \) and \( j \in \{1, \ldots, t\} \). Measure the output qubit for each execution, and let the result of the measurement for execution \((i, j)\) be \( y_{i,j} \in \{0, 1\} \).

2. For each \( i = 1, \ldots, s \), set
   \[
   z_i = \begin{cases} 
   1 & \text{if } \sum_{j=1}^{t} y_{i,j} \geq t \cdot \frac{a+b}{2} \\
   0 & \text{otherwise} 
   \end{cases}
   \]

3. Accept if \( \bigwedge_{i=1}^{s} z_i = 1 \), reject otherwise.

Now let us consider the maximum probability with which \( V' \) can be made to accept. Suppose first that an input \( x \in A_{\text{yes}} \) is fixed, so that \( V \) is made to accept with probability at least \( a \) by the prover \( P \). Our goal is to define a prover \( P' \) that causes \( V' \) to accept with probability at least \( 1 - 2^{-r} \). This is easily done by defining \( P' \) so that it runs \( st \) independent simulations of \( P \). Let \( Y_{i,j} \) and \( Z_i \), for \( i \in \{1, \ldots, s\} \) and \( j \in \{1, \ldots, t\} \), be Boolean-valued random variables corresponding to the values \( y_{i,j} \) and \( z_i \) when \( V' \) interacts with the prover \( P' \) just described. Given that \( P' \) simulates
st independent copies of $P$, we have that the random variables $Y_{i,j}$ are independent and satisfy $E[Y_{i,j}] \geq a$ for each pair $(i,j)$. By the Chernoff Bound, we therefore have

$$\Pr[Z_i = 0] = \Pr \left[ Y_{i,1} + \cdots + Y_{i,t} < a t \left( 1 - \frac{a - b}{2a} \right) \right] \leq \exp \left( -\frac{t}{8a} (a-b)^2 \right) \leq e^{-rs},$$

and thus the probability of rejection is at most $se^{-rs} < 2^{-r}$.

Suppose on the other hand that an input $x \in A_n$ is fixed, so that no prover $P$ can convince $V$ to accept with probability greater than $b$. Fix an arbitrary prover $P'$, and as before let $Y_{i,j}$ and $Z_i$ be Boolean-valued random variables corresponding to $y_{i,j}$ and $z_i$. There may not be independence among these random variables, as $P'$ may not treat the parallel executions independently. We do know, however, that $E[Y_{i,j}] \leq b$ for every $i, j$, given that the maximum acceptance probability of $V$ is $b$. By Markov’s Inequality we therefore have

$$\Pr[Z_i = 1] = \Pr \left[ Y_{i,1} + \cdots + Y_{i,t} \geq \frac{t(a + b)}{2} \right] \leq 2 \frac{E[Y_{i,1} + \cdots + Y_{i,t}]}{t(a + b)} < 1 - \frac{1}{2q}.$$ 

We may view $z_1, \ldots, z_s$ as being the outcomes of $s$ parallel executions of a quantum interactive proof system that accepts with probability at most $1 - \frac{1}{2q}$. The verifier $V'$ accepts if and only if all of these executions accept, and so by the result on parallel repetition proved in [KW00] we may conclude that the probability of acceptance of $V'$ is at most

$$\left( 1 - \frac{1}{2q} \right)^s < \exp \left( -\frac{s}{2q} \right) < 2^{-r}.$$ 

Thus, the verifier $V'$ has been shown to have completeness and soundness probabilities as required, completing the proof.

### 4 Maximum acceptance probability as a semidefinite program

The maximum acceptance probability of a verifier in a quantum interactive proof system can be phrased as semidefinite programming problem [KW00]. For this paper a semidefinite programming formulation based on ones described in [GW07, Wat09b] will be used.

Suppose $V$ is a two-message verifier, and that an input string $x$ of length $n$ is being considered. Let us also define

$$|\psi\rangle = U_x |0^{2^n}\rangle \quad \text{and} \quad \Pi = V_x^s (|1\rangle \langle 1| \otimes \mathbb{1}) V_x.$$ 

In words, $|\psi\rangle$ denotes the quantum state initially prepared by the verifier, the first half of which is sent to the prover; and $\Pi$ denotes the projection operator corresponding to the accept outcome of the measurement that the verifier effectively performs after receiving the prover’s message.

For convenience, we will assign distinct names to the complex vector spaces that arise from an interaction between $V$ on input $x$ and a compatible prover operation $\Psi$. Specifically, let $\mathcal{X}$ denote the space corresponding to the verifier’s message to the prover, let $\mathcal{Y}$ denote the space corresponding to the prover’s response, and let $\mathcal{Z}$ denote the space corresponding to the verifier’s private qubits. Thus, it holds that $|\psi\rangle \in \mathcal{X} \otimes \mathcal{Z}$ and $\Pi$ is a projection on $\mathcal{Y} \otimes \mathcal{Z}$. When the prover applies the operation $\Psi : L(\mathcal{X}) \rightarrow L(\mathcal{Y})$, the verifier accepts with probability

$$\left\langle \Pi, (\Psi \otimes \mathbb{1}_{L(\mathcal{Z})})(|\psi\rangle \langle \psi|) \right\rangle.$$  

(1)
To express the maximum probability for $V$ to accept, over all choices of an operation $\Psi$, as a semidefinite program, it is helpful to recall the Choi-Jamiołkowski representation of super-operators. Let us take $\{|0\rangle, \ldots, |N-1\rangle\}$ to be the standard basis of $\mathcal{X}$. Then the Choi-Jamiołkowski representation of $\Psi : L(\mathcal{X}) \to L(\mathcal{Y})$ is the operator $J(\Psi) \in L(\mathcal{Y} \otimes \mathcal{X})$ defined by
\[
J(\Psi) = \sum_{0 \leq i,j \leq N-1} \Psi(|i\rangle \langle j|) \otimes |i\rangle \langle j|.
\]
It holds that $\Psi$ is completely positive if and only if $J(\Psi)$ is positive semidefinite $[\text{Jam72}, \text{Cho75}]$, and $\Psi$ is trace-preserving if and only if $\text{Tr}_Y(J(\Psi)) = 1_X$.

Now let us write
\[
|\psi\rangle = \sum_{0 \leq i \leq N-1} |i\rangle |\psi_i\rangle
\]
for vectors $|\psi_0\rangle, \ldots, |\psi_{N-1}\rangle \in \mathcal{Z}$, and define $B \in L(\mathcal{X} \otimes \mathcal{Z})$ as
\[
B = \sum_{0 \leq i \leq N-1} |\psi_i\rangle \langle i|.
\]
Then it is clear that
\[
(\Psi \otimes 1_{L(\mathcal{Z})})(|\psi\rangle \langle \psi|) = (1_Y \otimes B)J(\Psi)(1_Y \otimes B^*)
\]
We therefore find that the probability of acceptance $\mu(Q)$ may alternately be written
\[
\langle \Pi(1_Y \otimes B)J(\Psi)(1_Y \otimes B^*) \rangle = \langle (1_Y \otimes B^*)\Pi(1_Y \otimes B), J(\Psi) \rangle = \langle Q, J(\Psi) \rangle
\]
for $Q = (1_Y \otimes B^*)\Pi(1_Y \otimes B)$. We call $Q$ the interactive measurement operator that is determined by $V$ on input $x$. It is clear that the interactive measurement operator $Q$ is positive semidefinite, and moreover that $Q \leq 1_Y \otimes \tilde{\xi}$ for the density operator $\tilde{\xi} = B^*B$.

Now, let us define
\[
\mu(Q) = \max_{\Psi} \langle Q, J(\Psi) \rangle,
\]
where the maximum is over all valid quantum operations of the form $\Psi : L(\mathcal{X}) \to L(\mathcal{Y})$. The quantity $\mu(Q)$ will be called the maximum acceptance probability of $Q$, as this value is precisely the maximum acceptance probability of the verifier $V$ on input $x$, whose description alone has led us to the definition of $Q$. As stated above, when $\Psi : L(\mathcal{X}) \to L(\mathcal{Y})$ ranges over the set of all valid quantum operations, $f(\Psi)$ ranges over the set of positive semidefinite operators satisfying the linear constraint $\text{Tr}_Y(J(\Psi)) = 1_X$. This implies that the quantity $\mu(Q)$ is represented by a semidefinite program:
\[
\begin{align*}
\text{maximize:} & \quad \langle Q, X \rangle \\
\text{subject to:} & \quad \text{Tr}_Y(X) \leq 1_X, \\
& \quad X \in \text{Pos}(\mathcal{Y} \otimes \mathcal{X}).
\end{align*}
\]
The feasible region of this semidefinite program is well-bounded (in the sense of [GLS93]), and therefore its optimal value $\mu(Q)$ can be approximated to high precision in time polynomial in the size of $Q$ (which is exponential in $|x|$). This fact does not help us to prove QIP(2) $\subseteq$ PSPACE, however. As is described in the next section, we will need an NC algorithm rather than just a polynomial-time algorithm to draw this conclusion.

It will be necessary for us to rephrase the semidefinite program above, and to explicitly state its dual program. As will be discussed shortly, we will only need to consider this formulation for
invertible interactive measurement operators, so \( Q \) is hereafter assumed to be invertible. Define a super-operator \( \Phi : L(Y \otimes X) \rightarrow L(X) \) as
\[
\Phi(X) = \text{Tr}_Y \left( Q^{-1/2} X Q^{-1/2} \right).
\]
The adjoint super-operator \( \Phi^* : L(X) \rightarrow L(Y \otimes X) \) to \( \Phi \) is given by
\[
\Phi^*(Y) = Q^{-1/2}(\mathbb{1}_Y \otimes Y)Q^{-1/2}.
\]
The value \( \mu(Q) \) is then seen to be the optimal value of the following semidefinite program:

**Primal problem**

- maximize: \( \text{Tr}(X) \)
- subject to: \( \Phi(X) \leq \mathbb{1}_X \), \( X \in \text{Pos}(Y \otimes X) \).

**Dual problem**

- minimize: \( \text{Tr}(Y) \)
- subject to: \( \Phi^*(Y) \geq \mathbb{1}_{Y \otimes X} \), \( Y \in \text{Pos}(X) \).

Strong duality follows from strict feasibility, which is easily verified, and so the optimal primal and dual solutions are given by \( \mu(Q) \).

## 5 Overview of the simulation

We will now explain, in high-level terms, our simulation of QIP(2) in PSPACE. To prove that QIP(2) \( \subseteq \) PSPACE, it will suffice to prove QIP(2) \( \subseteq \) NC(poly). This will be facilitated by the fact, discussed in Section 2.2, that many computations involving matrices, including elementary operations such as addition, multiplication, and inversion, as well as approximations of spectral decompositions, singular-value decompositions, and matrix exponentials, can be performed in NC.

For the remainder of this paper, assume that \( A = (A_{\text{yes}}, A_{\text{no}}) \) is an arbitrary promise problem in QIP(2), and let \( V \) be a two-message verifier for \( A \) that has exponentially small completeness and soundness error. The goal of the simulation is to determine whether or not \( V \) can be made to accept a given input string \( x \) with high probability. The variable \( n \) will always denote the input length \( n = |x| \), and \( p(n) \) will denote the number of qubits exchanged by the verifier and prover on each of the two messages as discussed in Section 3.

There are three main steps of the simulation:

1. Compute from \( x \) an explicit description of \( |\psi\rangle \) and \( \Pi \).
2. Process the description of the vector \( |\psi\rangle \) and the projection \( \Pi \) into a well-conditioned interactive measurement operator \( Q \) and positive rational numbers \( \gamma \) and \( \epsilon \) satisfying
   \[
   x \in A_{\text{yes}} \Rightarrow \mu(Q) \geq (1 + 4\epsilon)\gamma,
   \]
   \[
   x \in A_{\text{no}} \Rightarrow \mu(Q) \leq (1 - 4\epsilon)\gamma.
   \]
   For some polynomial \( q \) it will hold that \( \kappa(Q) \leq q(n) \), where \( \kappa(Q) = \|Q\| \|Q^{-1}\| \) denotes the condition number of \( Q \). Moreover it will hold that \( 1/q(n) \leq \epsilon \) and \( 1/q(n) \leq \gamma \).
3. Use a parallel algorithm, based on the multiplicative weights update method, to test whether \( \mu(Q) \) is larger or smaller than \( \gamma \).
The first step is easily performed in $\text{NC}(\text{poly})$, using an exact computation. In particular, one may simply compute products of the matrices that describe the individual gates of the verifier’s circuits. Given that this step is straightforward, we will not comment on it further. The second and third steps are more complicated, and are described separately in Sections 6 and 7 below. Both correspond to $\text{NC}$ computations (where the input size is exponential in $n$), and by composing these computations with the first step just described, we will obtain that $A$ is in $\text{NC}(\text{poly})$, and therefore $\text{QIP}(2) \subseteq \text{PSPACE}$.

### 6 Preparing a well-conditioned interactive measurement operator

After the first step of the simulation, we have a unit vector $|\psi\rangle$ and a projection operator $\Pi$. Let us write $M = 2^{p(n)}$ to denote the dimension of both of the message spaces and the verifier’s private work space defined by $V$ on input $x$, and let us also define $X_0 = \mathbb{C}^M$, $Y = \mathbb{C}^M$, and $Z_0 = \mathbb{C}^M$. We view that the space $X_0$ corresponds to the verifier’s message to the prover, that $Y$ corresponds to the prover’s message back to the verifier, and that $Z_0$ represents the verifier’s private workspace; and thus $|\psi\rangle \in X_0 \otimes Z_0$ and $\Pi \in \text{Pos}(Y \otimes Z_0)$. The reason for subscripting $X_0$ and $Z_0$ with 0 is that we are viewing these as initial choices of spaces. In the processing of $|\psi\rangle$ and $\Pi$, we will define an interactive measurement operator $Q$ over spaces $X$, $Y$ and $Z$ where $X = \mathbb{C}^N$ and $Z = \mathbb{C}^N$ for some choice of a positive integer $N \leq M$.

Along the same lines as was discussed in Section 3, we may define an interactive measurement operator $R \in \text{Pos}(Y \otimes X')$ as $R = (\mathbb{1}_Y \otimes B^*) \Pi (\mathbb{1}_Y \otimes B)$, for

$$|\psi\rangle = \sum_{0 \leq i \leq M-1} |i\rangle |\psi_i\rangle \quad \text{and} \quad B = \sum_{0 \leq i \leq M-1} |\psi_i\rangle \langle i|.$$ 

The quantity $\mu(R)$ is precisely the maximum acceptance probability of $V$ on input $x$, but nothing can be said about the condition number of $R$ (which may not even be invertible).

Our goal is to compute a new measurement operator $Q \in \text{Pos}(Y \otimes X')$, where $X' = \mathbb{C}^N$ for some choice of $N \leq M$, along with positive rational numbers $\gamma$ and $\epsilon$, such that the following properties hold for some polynomial $q$:

1. The interactive measurement operator $Q$ is well-conditioned: $\kappa(Q) \leq q(n)$.
2. The values $\gamma$ and $\epsilon$ are non-negligible: $1/q(n) \leq \epsilon$ and $1/q(n) \leq \gamma$.
3. The value $\mu(Q)$ satisfies the properties

   $$x \in A_{\text{yes}} \Rightarrow \mu(Q) \geq (1 + 4\epsilon)\gamma,$$
   $$x \in A_{\text{no}} \Rightarrow \mu(Q) \leq (1 - 4\epsilon)\gamma.$$  

The first step in this process is to replace $|\psi\rangle$ by a more uniform vector $|\phi\rangle \in X_0 \otimes Z_0$ that is “similar enough” to $|\psi\rangle$ in the sense to be described. We will, in particular, take $|\phi\rangle$ to be maximally entangled over certain subspaces of $X_0$ and $Z_0$. This is done by performing the following operations:

1. Let

   $$|\psi\rangle = \sum_{0 \leq j \leq M-1} \sqrt{\lambda_j} |x_j\rangle |z_j\rangle$$

   be a Schmidt decomposition of $|\psi\rangle$. 

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2. For each positive integer $i$, define the interval $I_i = \left(2^{-i}, 2^{-i+1}\right]$, and define

$$\Sigma_i = \{ j \in \{0, \ldots, M - 1\} : \lambda_j \in I_i \}.$$ 

3. Let $k = p(n) + 1$ and choose $i \in \{1, \ldots, k\}$ so that

$$\sum_{j \in \Sigma_i} \lambda_j \geq \frac{1}{2k}.$$ 

The fact that such an $i$ exists is proved below, and hereafter we write $\Sigma = \Sigma_i$ for this choice of $i$. 

4. Define

$$|\phi\rangle = \frac{1}{\sqrt{|\Sigma|}} \sum_{j \in \Sigma} |x_j \rangle |z_j \rangle.$$ 

Now, consider the interactive measurement operator $S \in \text{Pos}(\mathcal{Y} \otimes \mathcal{X}_0)$ that is obtained by replacing $|\psi\rangle$ with $|\phi\rangle$ (with $\Pi$ unchanged). In other words, $S$ is defined by the same process as $R$ (which was determined by $|\psi\rangle$ and $\Pi$ as described above), and satisfies the equation

$$\langle S, J(\Psi) \rangle = \left\langle \Pi, (\Psi \otimes 1_{L(Z_0)})(|\phi\rangle\langle\phi|) \right\rangle$$

for every super-operator $\Psi : L(\mathcal{X}_0) \rightarrow L(\mathcal{Y})$. We will prove that

$$\mu(R) - \left(1 - \frac{1}{8k}\right) \leq \mu(S) \leq 4k \mu(R). \quad (3)$$

These are fairly loose bounds—but for the two extremes where $\mu(R)$ is exponentially close to 0 or 1, the corresponding values for $\mu(S)$ will be separated by the reciprocal of a polynomial, which is good enough for our needs.

First, note that

$$\sum_{i > k} \sum_{j \in \Sigma_i} \lambda_j \leq M 2^{-k} \leq \frac{1}{2}$$

and therefore

$$\sum_{i=1}^{k} \sum_{j \in \Sigma_i} \lambda_j \geq \frac{1}{2}.$$ 

Thus, there must exist a suitable choice of $i$ in step 3 so that

$$\sum_{j \in \Sigma_i} \lambda_j \geq \frac{1}{2k}.$$ 

as claimed. A lower bound on the size of $\Sigma = \Sigma_i$ may be obtained by noting that

$$\sum_{j \in \Sigma} \lambda_j \leq 2^{-i+1} |\Sigma|$$

and therefore

$$|\Sigma| \geq \frac{2^i}{4k}.$$
Now, the inner product of $|\psi\rangle$ and $|\phi\rangle$ is easily bounded from below as

$$\langle \phi | \psi \rangle = \sum_{j \in \Sigma} \sqrt{\frac{\lambda_j}{|\Sigma|}} \geq |\Sigma| \sqrt{\frac{2^{-i}}{|\Sigma|}} = \sqrt{|\Sigma|} 2^{-i} \geq \frac{1}{\sqrt{4k}},$$

and therefore

$$\| |\psi\rangle \langle \psi| - |\phi\rangle \langle \phi| \|_1 = 2\sqrt{1 - |\langle \phi | \psi \rangle|^2} \leq 2\sqrt{1 - \frac{1}{4k}} \leq 2 \left( 1 - \frac{1}{8k} \right).$$

For every choice of an admissible super-operator $\Psi : L(\mathcal{X}_0) \to L(\mathcal{Y})$ it holds that

$$0 \leq \left( \Psi^* \otimes 1_{L(\mathcal{Z}_0)} \right) (\Pi) \leq 1_{\mathcal{X}_0 \otimes \mathcal{Z}_0},$$

and by combining this observation with the fact that $|\psi\rangle \langle \psi| - |\phi\rangle \langle \phi|$ is traceless we obtain

$$\langle R - S, J(\Psi) \rangle = \left( \left( \Psi^* \otimes 1_{L(\mathcal{Z}_0)} \right) (\Pi), |\psi\rangle \langle \psi| - |\phi\rangle \langle \phi| \right) \leq \frac{1}{2} \| |\psi\rangle \langle \psi| - |\phi\rangle \langle \phi| \|_1 \leq 1 - \frac{1}{8k}.$$ Therefore

$$\mu(R) - \mu(S) \leq 1 - \frac{1}{8k},$$

which establishes the lower bound on $\mu(S)$ claimed in (3) above.

To establish the upper bound on $\mu(S)$, which is the second inequality in (3), let us first choose an admissible super-operator $\Psi$ so that

$$\mu(S) = \left( \Pi, \left( \Psi \otimes 1_{L(\mathcal{Z}_0)} \right) (|\phi\rangle \langle \phi|) \right).$$

Now, observe that

$$\frac{1}{4k} \frac{1}{|\Sigma|} \leq 2^{-i} < \lambda_j$$

for each $j \in \Sigma$, and thus

$$\frac{1}{4k} \text{Tr}_{\mathcal{X}_0}(|\phi\rangle \langle \phi|) \leq \text{Tr}_{\mathcal{X}_0}(|\psi\rangle \langle \psi|).$$

It is therefore possible to choose a density operator $\xi \in D(\mathcal{Y} \otimes \mathcal{Z}_0)$ such that

$$\left( 1 - \frac{1}{4k} \right) \text{Tr}_{\mathcal{Y}}(\xi) = \text{Tr}_{\mathcal{X}_0}(|\psi\rangle \langle \psi|) - \frac{1}{4k} \text{Tr}_{\mathcal{X}_0}(|\phi\rangle \langle \phi|).$$

Because $\Psi$ is admissible, we may therefore conclude that

$$\text{Tr}_{\mathcal{X}_0}(|\psi\rangle \langle \psi|) = \text{Tr}_{\mathcal{Y}} \left( \frac{1}{4k} \left( \Psi \otimes 1_{L(\mathcal{Z}_0)} \right) (|\phi\rangle \langle \phi|) + \left( 1 - \frac{1}{4k} \right) \xi \right),$$

and so there must exist an admissible super-operator $\Xi : L(\mathcal{X}_0) \to L(\mathcal{Y})$ so that

$$\left( \Xi \otimes 1_{L(\mathcal{Z}_0)} \right) (|\psi\rangle \langle \psi|) = \frac{1}{4k} \left( \Psi \otimes 1_{L(\mathcal{Z}_0)} \right) (|\phi\rangle \langle \phi|) + \left( 1 - \frac{1}{4k} \right) \xi.$$ Consequently we have

$$\mu(R) \geq \langle R, J(\Xi) \rangle = \left( \Pi, \left( \Xi \otimes 1_{L(\mathcal{Z}_0)} \right) (|\psi\rangle \langle \psi|) \right) \geq \frac{1}{4k} \left( \Pi, \left( \Psi \otimes 1_{L(\mathcal{Z}_0)} \right) (|\phi\rangle \langle \phi|) \right) = \frac{1}{4k} \mu(S)$$
as required.

Having obtained a uniform vector $|\phi\rangle$ that gives (when combined with $\Pi$) an interactive measurement operator $S$ satisfying (3), we must make a couple of additional modifications to be sure that a well-conditioned interactive measurement operator has been obtained.

First, we will replace $X_0$ and $Z_0$ with the spaces $X = C^N$ and $Z = C^N$ for $N = |\Sigma|$. Let $X \in U(X, X_0)$ and $Z \in U(Z, Z_0)$ be linear isometries defined as

$$X = \sum_{j=1}^{N} |x_j\rangle\langle j| \quad \text{and} \quad Z = \sum_{j=1}^{N} |z_j\rangle\langle j|,$$

and define

$$|\tau\rangle = (X^* \otimes Z^*) |\phi\rangle = \frac{1}{\sqrt{N}} \sum_{j=1}^{N} |j\rangle |j\rangle \quad \text{and} \quad P = (1_y \otimes 1_x) \Pi (1_y \otimes Z).$$

It is clear that $|\tau\rangle$ is a unit vector and $P$ is an ordinary measurement operator. (It might not be that $P$ is a projection operator, but it is positive semidefinite and satisfies $P \leq 1_y \otimes Z$.) Finally, let $Q \in \text{Pos}(Y \otimes X)$ be the interactive measurement operator defined by the vector $|\tau\rangle$ and the ordinary measurement operator

$$\left( 1 - \frac{1}{64k} \right) P + \frac{1}{64k} 1_{Y \otimes Z}.$$

It holds that

$$\mu(Q) = \left( 1 - \frac{1}{64k} \right) \mu(S) + \frac{1}{64k},$$

and therefore

$$\mu(R) - \left( 1 - \frac{1}{8k} \right) \leq \mu(Q) \leq 4k\mu(R) + \frac{1}{64k}. \quad (4)$$

Now let us verify that $Q$ has the properties we require of it. First let us consider the condition number $\kappa(Q)$. It is easily shown that

$$Q \geq \frac{1}{64kN} 1_{Y \otimes X} \quad \text{and} \quad Q \leq \frac{1}{N} 1_{Y \otimes X},$$

and therefore $\kappa(Q) \leq 64k$. It remains to define nonnegligible values $\gamma$ and $\epsilon$, and to consider their relationship to $\mu(Q)$ in the two cases: $x \in A_{\text{yes}}$ and $x \in A_{\text{no}}$.

We have assumed that the original interactive proof system has exponentially small completeness and soundness errors, and therefore we may assume that for sufficiently large $n$ we have

$$x \in A_{\text{no}} \Rightarrow \mu(R) \leq \frac{1}{256k^2} \quad \text{and} \quad x \in A_{\text{yes}} \Rightarrow \mu(R) \geq 1 - \frac{1}{32k}.$$

(Alternately we may assume these bounds hold for all $n$ by hard-coding small inputs $x$ into the verifier.) Thus, by the bounds (4) above, we have

$$x \in A_{\text{no}} \Rightarrow \mu(Q) \leq \frac{1}{32k} \quad \text{and} \quad x \in A_{\text{yes}} \Rightarrow \mu(Q) \geq \frac{1}{16k}.$$

By taking

$$\gamma = \frac{3}{64k} = \frac{3}{64(p(n) + 1)}, \quad \epsilon = \frac{1}{12}, \quad \text{and} \quad q(n) \geq 64k = 64(p(n) + 1),$$

we therefore have the properties required.
1. Let 
\[ \delta = \frac{\varepsilon^2}{8 \kappa(Q)^2} \quad \text{and} \quad T = \left\lceil \frac{24 \ln(NM)}{\varepsilon^3 \gamma^3 \delta} \right\rceil \].

2. Let \( W_0 = \mathbb{1}_Y \otimes X \) and let \( \rho_0 = W_0 / \text{Tr}(W_0) \).

3. For \( t = 0, \ldots, T - 1 \) do:
   a. Compute a spectral decomposition of \( \Phi(\rho_t) \):
      \[ \Phi(\rho_t) = \sum_{j=1}^{N} \lambda_j |x_j \rangle \langle x_j| . \]
   b. Let \( S = \{ j \in \{1, \ldots, N\} : \gamma \lambda_j > 1 \} \) and let \( s = \sum_{j \in S} \lambda_j \).
   c. If \( s \leq \delta \|Q^{-1}\| \), then halt and accept.
   d. Let 
      \[ Y_t = \frac{1}{s} \sum_{j \in S} |x_j \rangle \langle x_j| . \]
   e. Let 
      \[ W_{t+1} = \exp \left( -\frac{\varepsilon \gamma \delta}{2} \Phi^*(Y_0 + \cdots + Y_t) \right) , \]
      and let \( \rho_{t+1} = W_{t+1} / \text{Tr}(W_{t+1}) \).

4. Halt and reject.

Figure 2: Algorithm to test if \( \mu(Q) \geq \gamma \).

7 Verifying maximum acceptance probability

We now describe and analyze a parallel algorithm, based on the multiplicative weights update method, to distinguish the two cases (2) from the previous section. The algorithm operates as described in Figure 2 and the super-operators \( \Phi \) and \( \Phi^* \) are as defined in Section 4.

7.1 Lemmas used in the analysis

We will need a few basic facts and three lemmas to analyze the algorithm described in Figure 2. We begin by noting two facts concerning matrix exponentials. First, the Golden-Thompson Inequality (see Section IX.3 of [Bha97]) states that, for any two Hermitian matrices \( X \) and \( Y \) of equal dimension, we have
\[ \text{Tr} \left( e^{X+Y} \right) \leq \text{Tr} \left( e^X e^Y \right) . \]
Second is the following inequality concerning the matrix exponential of positive semidefinite matrices.

**Lemma 7.1.** Let \( P \) be an operator satisfying \( 0 \leq P \leq 1 \). Then for every real number \( \eta > 0 \), it holds that
\[ \exp(-\eta P) \leq 1 - \eta \exp(-\eta) P . \]
Lemma 7.4. It is sufficient to prove the inequality for $P$ replaced by a scalar $\lambda \in [0, 1]$, for then the operator inequality follows by considering a spectral decomposition of $P$. If $\lambda = 0$ then the inequality is trivial, so assume $\lambda > 0$. By the Mean Value Theorem there exists a value $\lambda_0 \in (0, \lambda)$ such that

$$\frac{\exp(-\eta \lambda) - 1}{\lambda} = -\eta \exp(-\eta \lambda_0) \leq -\eta \exp(-\eta),$$

which yields the inequality.

For the proofs of the remaining two lemmas, some additional notation is used. Suppose that $\mathcal{X} = \mathbb{C}^N$ and $\mathcal{Y} = \mathbb{C}^M$, and assume that the standard bases of these spaces are $\{|0\rangle, \ldots, |N-1\rangle\}$ and $\{|0\rangle, \ldots, |M-1\rangle\}$, respectively. Then we define a linear mapping

$$\vec : L(\mathcal{X}, \mathcal{Y}) \rightarrow \mathcal{Y} \otimes \mathcal{X}$$

by taking $\vec(|i\rangle \langle j|) = |i\rangle |j\rangle$ for each choice of $i \in \{0, \ldots, M-1\}$ and $j \in \{0, \ldots, N-1\}$.

Lemma 7.2. Let $\mathcal{X} = \mathbb{C}^N$ and $\mathcal{Y} = \mathbb{C}^M$ for positive integers $N$ and $M$. Let $P_0, P_1 \in \text{Pos}(\mathcal{X})$ and let $R_0 \in \text{Pos}(\mathcal{X} \otimes \mathcal{Y})$ satisfy $\text{Tr}_\mathcal{Y}(R_0) = P_0$. Then there exists an operator $R_1 \in \text{Pos}(\mathcal{X} \otimes \mathcal{Y})$ such that $\text{Tr}_\mathcal{Y}(R_1) = P_1$ and $F(R_0, R_1) = F(P_0, P_1)$.

Proof. By the monotonicity of the fidelity function, it must hold that $F(R_0, R_1) \leq F(P_0, P_1)$ for every choice of $R_1$ satisfying $\text{Tr}_\mathcal{Y}(R_1) = P_1$. It therefore suffices to show that equality can be achieved.

Choose a unitary operator $V \in U(\mathcal{X})$ for which $\sqrt{P_0} \sqrt{P_1} V$ is positive semidefinite. For such a $V$ it holds that $F(P_0, P_1) = \text{Tr}(\sqrt{P_0} \sqrt{P_1} V)$. Now let $\mathcal{W} = \mathbb{C}^{NM}$ and let $|u_0\rangle \in \mathcal{Y} \otimes \mathcal{X} \otimes \mathcal{W}$ be a purification of $R_0$. Given that $|u_0\rangle$ also purifies $P_0$, it must take the form

$$|u_0\rangle = \vec\left(\sqrt{P_0} U^*\right)$$

for some choice of a linear isometry $U \in U(\mathcal{X}, \mathcal{Y} \otimes \mathcal{W})$. Finally, let

$$R_1 = \text{Tr}_\mathcal{W}\left(\vec\left(\sqrt{P_1} V U^*\right) \vec\left(\sqrt{P_1} V U^*\right)^*\right).$$

It holds that

$$F(R_0, R_1) \geq \left|\left\langle \sqrt{P_0} U^*, \sqrt{P_1} V U^*\right\rangle\right| = \text{Tr}\left(\sqrt{P_0} \sqrt{P_1} V\right) = F(P_0, P_1)$$

as required.

Remark 7.3. Lemma 7.2 is a fairly straightforward extension of Uhlmann’s Theorem [Uhl76]. (See also pages 410–411 of [NC00].)

Lemma 7.4. Let $R_0, R_1 \in \text{Pos}(\mathcal{W})$ for $\mathcal{W} = \mathbb{C}^k$. Then

$$\|R_0 - R_1\|_1 \leq \sqrt{2 \text{Tr}(R_0)^2 + 2 \text{Tr}(R_1)^2 - 4 F(R_0, R_1)^2}.$$

Proof. Choose $V \in U(\mathcal{W})$ so that $\sqrt{R_0} \sqrt{R_1} V$ is positive semidefinite, and therefore $F(R_0, R_1) = \text{Tr}(\sqrt{R_0} \sqrt{R_1} V)$. We have that

$$\vec\left(\sqrt{R_0}\right) \in \mathcal{W} \otimes \mathcal{W} \quad \text{and} \quad \vec\left(\sqrt{R_1} V\right) \in \mathcal{W} \otimes \mathcal{W}$$

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purify \( R_0 \) and \( R_1 \), respectively, so by the monotonicity of the trace norm it holds that
\[
\| R_0 - R_1 \|_1 \leq \| \text{vec} \left( \sqrt{R_0} \right) \text{vec} \left( \sqrt{R_0} \right)^* - \text{vec} \left( \sqrt{R_1} V \right) \text{vec} \left( \sqrt{R_1} V \right)^* \|_1.
\]
Using the inequality
\[
\| A \|_1 \leq \sqrt{\text{rank}(A)} \| A \|_2,
\]
which follows easily from the expressions of the trace and Frobenius norms in terms of singular values, along with the Cauchy–Schwarz inequality, we have
\[
\| R_0 - R_1 \|_1 \leq \sqrt{2} \| \text{vec} \left( \sqrt{R_0} \right) \text{vec} \left( \sqrt{R_0} \right)^* - \text{vec} \left( \sqrt{R_1} V \right) \text{vec} \left( \sqrt{R_1} V \right)^* \|_2
\]
\[
= \sqrt{2 \text{Tr}(R_0)^2 + 2 \text{Tr}(R_1)^2 - 4 \text{Tr} \left( \sqrt{R_0} \sqrt{R_1} V \right)^2}
\]
\[
= \sqrt{2 \text{Tr}(R_0)^2 + 2 \text{Tr}(R_1)^2 - 4 F(R_0, R_1)^2}
\]
as required.

**Remark 7.5.** When \( R_0 = \rho_0 \) and \( R_1 = \rho_1 \) for density operators \( \rho_0 \) and \( \rho_1 \), we obtain the familiar inequality
\[
\| \rho_0 - \rho_1 \|_1 \leq 2 \sqrt{1 - F(\rho_0, \rho_1)^2},
\]
or equivalently
\[
F(\rho_0, \rho_1) \leq \sqrt{1 - \frac{1}{4} \| \rho_0 - \rho_1 \|_1^2},
\]
which is one of the Fuchs-van de Graaf inequalities [FvdG99].

### 7.2 Analysis of the algorithm (ignoring precision)

Our algorithm cannot be implemented exactly using bounded-depth Boolean circuits: the spectral decompositions and matrix exponentials can only be approximated. However, for the sake of exposition, the issue of precision will be completely ignored in this subsection; meaning that we will imagine that all of the operations can be performed exactly. In the section that follows this one, the actual precision requirements of the algorithm are considered. As is shown there, it turns out that the algorithm is not particularly sensitive to errors, and in fact it is possible to perform all of the required computations in parallel with exponentially greater precision than would be required for the correctness of the algorithm.

Let us consider first the case that the algorithm accepts. Let \( \rho = \rho_1 \) for the iteration \( t \) of the loop in step 3 in which acceptance occurs. To prove that the algorithm has answered correctly, we will construct an operator \( X \in \text{Pos}(Y \otimes X) \) such that \( \Phi(X) \leq \mathbb{1}_Y \) and \( \text{Tr}(X) \geq (1 - \varepsilon)\gamma \); and therefore \( \mu(Q) \geq (1 - \varepsilon)\gamma \). By the conditions (2) on \( Q \), this implies that \( \mu(Q) \geq (1 + 4\varepsilon)\gamma \), and therefore \( x \in A_{\text{yes}} \), as required.

The operator \( X \) is defined as follows. First, let \( R_0 = Q^{-1/2} \rho Q^{-1/2} \), let \( P_0 = \text{Tr}_Y(R_0) = \Phi(\rho) \), and let
\[
P_1 = \frac{1}{\gamma} \sum_{j \in S} |x_j \langle x_j| + \sum_{j \notin S} \lambda_j |x_j \langle x_j|.
\]
By Lemma \[ \text{Lemma 7.2} \] there must exist \( R_1 \in \text{Pos}(\mathcal{Y} \otimes \mathcal{X}) \) such that \( \text{Tr}_Y(R_1) = P_1 \) and \( F(R_0, R_1) = F(P_0, P_1) \). We then take
\[
X = \gamma \sqrt{Q R_1 \sqrt{Q}}.
\]
It holds that \( X \geq 0 \) and \( \Phi(X) \leq 1, X \). To establish a lower bound on \( \text{Tr}(X) \), we first note that
\[
1 - \text{Tr} \left( \sqrt{Q R_1 \sqrt{Q}} \right) = \langle Q, R_0 - R_1 \rangle \leq \| Q \| \| R_0 - R_1 \|_1.
\]
By Lemma \[ \text{Lemma 7.4} \] we conclude that
\[
\| R_0 - R_1 \|_1 \leq \sqrt{2 \text{Tr}(P_0)^2 + 2 \text{Tr}(P_1)^2 - 4 F(P_0, P_1)^2}.
\]
It holds that \( \text{Tr}(P_1) \leq \text{Tr}(P_0) \), and we also have
\[
F(P_0, P_1) \geq \sum_{j \in S} \lambda_j = \text{Tr}(P_0) - s \geq \text{Tr}(P_0) - \delta \| Q^{-1} \|.
\]
Therefore, given that \( \text{Tr}(P_0) \leq \| Q^{-1} \| \), we have \( \| R_0 - R_1 \|_1 \leq \sqrt{8 \delta} \| Q^{-1} \| \), and so
\[
1 - \text{Tr} \left( \sqrt{Q R_1 \sqrt{Q}} \right) \leq \sqrt{8 \delta} \| Q^{-1} \| \| Q \| = \sqrt{8 \delta} \kappa(Q) = \varepsilon.
\]
It follows that
\[
\text{Tr}(X) = \gamma \text{Tr} \left( \sqrt{Q R_1 \sqrt{Q}} \right) \geq (1 - \varepsilon) \gamma
\]
as required.

Now let us consider the case that the algorithm rejects. Along similar lines to the previous case, we will construct an operator \( Y \in \text{Pos}(\mathcal{X}) \) such that \( \Phi^+(Y) \geq 1_{\mathcal{Y} \otimes \mathcal{X}} \) and \( \text{Tr}(Y) \leq (1 + \varepsilon) \gamma \). By the conditions \[ \text{(2)} \] on \( Q \) this implies that \( \mu(Q) \leq (1 - 4\varepsilon) \gamma \), and therefore \( x \in A_{\text{no}} \). In particular, we may take
\[
Y = \frac{1 + \varepsilon}{t} (Y_0 + \cdots + Y_{T-1}).
\]
Each operator \( Y_t \) satisfies
\[
\text{Tr}(Y_t) = \frac{|S|}{s} < \frac{1}{s} \sum_{j \in S} \gamma \lambda_j = \gamma,
\]
and therefore \( \text{Tr}(Y) < (1 + \varepsilon) \gamma \). Each \( Y_t \) is also clearly positive semidefinite, so it remains to prove that \( \Phi^+(Y) \geq 1_{\mathcal{Y} \otimes \mathcal{X}} \). To this end we will first establish two conditions on each operator \( Y_t \). First, we have
\[
\langle \rho_t, \Phi^+(Y_t) \rangle = \langle \Phi(\rho_t), Y_t \rangle = \frac{1}{s} \sum_{j \in S} \lambda_j = 1.
\]
(5)

Second, given that \( s > \delta \| Q^{-1} \| \) for the case at hand, we have
\[
\| \Phi^+(Y_t) \| = \| Q^{-1/2}(1_{\mathcal{Y}} \otimes Y_t)Q^{-1/2} \| = \| Q^{-1} \| \| Y_t \| = \frac{\| Q^{-1} \|}{s} < \frac{1}{\delta},
\]
and therefore \( \| \delta \Phi^+(Y_t) \| < 1 \).

Now, for the sake of clarity, let us write \( \eta = \varepsilon \gamma / 2 \). Note that, for \( 0 \leq t \leq T - 1 \), it holds that
\[
\text{Tr}(W_{t+1}) = \text{Tr} [\exp(-\eta \delta \Phi^+(Y_0 + \cdots + Y_t))]
\leq \text{Tr} [\exp(-\eta \delta \Phi^+(Y_0 + \cdots + Y_{t-1})) \exp(-\eta \delta \Phi^+(Y_t))]
= \text{Tr} [W_t \exp(-\eta \delta \Phi^+(Y_t))],
\]
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where we have used the Golden–Thompson Inequality. Given that \( \| \delta \Phi^*(Y_t) \| \leq 1 \) we have by Lemma 7.1 that

\[
\exp(-\eta \delta \Phi^*(Y_t)) \leq 1 - \eta \delta \exp(-\eta) \Phi^*(Y_t),
\]

and therefore

\[
\text{Tr}(W_{t+1}) \leq \text{Tr}(W_t) (1 - \eta \delta \exp(-\eta) \langle \rho_t, \Phi^*(Y_t) \rangle) \leq \text{Tr}(W_t) \exp(-\eta \delta \exp(-\eta)).
\]

(Here we have used the inequality \( \exp(-\alpha) \geq 1 - \alpha \), which holds for all real numbers \( \alpha \), as well as the fact that \( \text{Tr}(AB) \leq \text{Tr}(AC) \) whenever \( A \geq 0 \) and \( B \leq C \).) Repeating this argument, and substituting \( \text{Tr}(W_0) = NM \), we have

\[
\text{Tr}(W_T) \leq NM \exp(-\eta \delta T \exp(-\eta)).
\]

On the other hand, it is clear that

\[
\text{Tr}(W_T) = \text{Tr}[\exp(-\eta \delta \Phi^*(Y_0 + \cdots + Y_{T-1}))] \geq \exp(-\eta \delta \lambda_{NM} \Phi^*(Y_0 + \cdots + Y_{T-1})),
\]

and therefore

\[
\lambda_{NM} \left( \Phi^* \left( \frac{Y_0 + \cdots + Y_{T-1}}{T} \right) \right) \geq \exp(-\eta) - \frac{\ln(NM)}{\eta \delta T}.
\]

Substituting the specified value of \( T \), and using the fact that \( \exp(-\eta) - \frac{\eta^2}{2} \geq 1 - \eta \) (which holds for any \( \eta \in [0, 1] \)), we have

\[
\lambda_{NM} \left( \Phi^* \left( \frac{Y_0 + \cdots + Y_{T-1}}{T} \right) \right) \geq 1 - \eta = 1 - \frac{\varepsilon \gamma}{2}.
\]

Therefore

\[
\lambda_{NM}(\Phi^*(Y)) \geq (1 + \varepsilon) \left( 1 - \frac{\varepsilon \gamma}{2} \right) \geq 1,
\]

and so \( \Phi^*(Y) \geq \mathbb{1}_{Y \otimes X} \) as required.

We therefore have that the algorithm works correctly, modulo the precision issues to be discussed in the section following this one. It remains only to observe that it can be implemented in NC (meaning that it results in an NC(poly) computation when composed with the first two steps of the simulation). Some of the details required to argue this can be found below; but at a high level one sees that each iteration of the loop in step 3 can be performed with high precision in NC, and the total number of iterations required is polynomial in \( n \) (and therefore polylogarithmic in the size of \( Q \)).

### 8 Precision requirements for the simulation

We now discuss the precision that is required for the simulation to yield a correct answer. It will turn out that the simulation is not particularly sensitive to errors, and one could in fact afford to take exponentially more precision than is required and still be within the class NC(poly).

The first step of the simulation, in which an explicit description of \( |\psi\rangle \) and \( \Pi \) is obtained, can be performed exactly in NC(poly) as has already been observed. So, let us move on to the second step, in which \( |\psi\rangle \) and \( \Pi \) are processed to obtain a well-conditioned interactive measurement operator \( Q \). This step requires the approximation of one singular value decomposition (to approximate the Schmidt decomposition of \( |\psi\rangle \)), along with a few other operations that can be performed exactly or with high precision in NC.
For the moment let us denote by $\tilde{Q}$ the actual operator that is computed by an NC implementation of this step, as opposed to the true operator $Q$ that would be output by an idealized, exact complex number algorithm. By computing the singular value decomposition to high precision, it is possible to take such an approximation so that

$$\|Q - \tilde{Q}\| < 2^{-\text{poly}(n)},$$

where $\text{poly}$ denotes any polynomial of our choice. It is not difficult to prove that the quantity $|\mu(Q) - \mu(\tilde{Q})|$ is upper-bounded by $N$ times $\|Q - \tilde{Q}\|$, and therefore we may take $\tilde{Q}$ so that

$$\left|\mu(Q) - \mu(\tilde{Q})\right| < 2^{-2\text{poly}(n)},$$

again for $\text{poly}$ denoting any polynomial of our choice. We do not need this much precision: we only need a $1/\text{poly}(n)$ separation between the values of $\mu(\tilde{Q})$ for the cases $x \in A_{\text{yes}}$ and $x \in A_{\text{no}}$, which requires that $\|Q - \tilde{Q}\|$ is exponentially (rather than double-exponentially) small in $n$. So, to be concrete, we may decide to take sufficient precision so that

$$\|Q - \tilde{Q}\| < \varepsilon \gamma / N,$$

and therefore

$$|\mu(Q) - \mu(\tilde{Q})| < \varepsilon \gamma.$$

Thus,

$$x \in A_{\text{yes}} \Rightarrow \mu(\tilde{Q}) \geq (1 + 3\varepsilon)\gamma$$

$$x \in A_{\text{no}} \Rightarrow \mu(\tilde{Q}) \leq (1 - 3\varepsilon)\gamma.$$

(6)

Hereafter we will return to writing $Q$ rather than $\tilde{Q}$, with the understanding that $Q$ now represents an approximation that is stored by our algorithm. In addition, we will assume that $\sqrt{Q}$, and therefore $Q^{-1/2}$ as well, has Gaussian rational entries and is known precisely. This assumption is easily met by replacing $Q$ with the square of a high precision approximation to $\sqrt{Q}$. The point of this assumption is that we avoid having to consider an additional error term every time $\sqrt{Q}$ or $Q^{-1/2}$ is involved in any computation. (There is no reason beyond simplifying the analysis to make this assumption.)

Now suppose that the algorithm described in Section 7 is performed with limited precision. Consider first the case that the algorithm accepts, and let $\rho = \rho_t$ denote the density operator that is stored by the algorithm on the iteration $t$ in which acceptance occurs. Note that it is not necessary to view that $\rho$ is an approximation of something else: the simple fact that $\rho$ causes acceptance will allow us to conclude that $x \in A_{\text{yes}}$ in a similar way to the error-free analysis.

Specifically, we will construct an operator $X \in \text{Pos}(\mathcal{Y} \otimes \mathcal{X})$ such that $\Phi(X) \leq 1_y$ and $\text{Tr}(X) \geq (1 - 2\varepsilon)\gamma$. As before we will let $R_0 = Q^{-1/2}\rho Q^{-1/2}$ and $P_0 = \text{Tr}_y(R_0)$. This time, we must consider that the spectral decomposition is approximate. Let us write

$$\tilde{P}_0 = \sum_{j=1}^N \lambda_j |x_j\rangle\langle x_j|$$

to denote the approximate value of the spectral decomposition, so that $\|P_0 - \tilde{P}_0\|$ represents the error in this approximation, and let us assume that sufficient accuracy is taken so that

$$\|P_0 - \tilde{P}_0\| < \frac{\delta}{4N \|Q^{-1}\|}.$$
As above, this is significantly less accuracy than is available—for we could take
\[ \| P_0 - \overline{P}_0 \| < 2^{-2\text{poly}(n)} \]
if it were advantageous. Continuing on as before, let
\[ P_1 = \frac{1}{\gamma} \sum_{j \in S} |x_j\rangle \langle x_j| + \sum_{j \not\in S} \lambda_j |x_j\rangle \langle x_j|, \]
let \( R_1 \in \text{Pos}(\mathcal{Y} \otimes \mathcal{X}) \) to be an extension of \( P_1 \) for which \( F(R_0, R_1) = F(P_0, P_1) \), and let
\[ X = \gamma \sqrt{Q} R_1 \sqrt{Q}. \]

We have that \( X \geq 0 \) and \( \Phi(X) \leq 1 \) as before, and to establish a lower bound on \( \text{Tr}(X) \) we again use the fact that
\[ 1 - \text{Tr} \left( \sqrt{Q} R_1 \sqrt{Q} \right) \leq \| Q \| \| R_0 - R_1 \|_1, \]
as well as the bound
\[ \| R_0 - R_1 \|_1 \leq \sqrt{2 \text{Tr}(P_0)^2 + 2 \text{Tr}(P_1)^2 - 4 F(P_0, P_1)^2}. \]

Based on the bound (7), it follows that
\[ \text{Tr}(P_1)^2 \leq \text{Tr}(P_0)^2 + \delta \]
and
\[ F(P_0, P_1)^2 \geq \left( \text{Tr}(P_0) - \delta \| Q^{-1} \| \right)^2 + \delta, \]
and therefore
\[ \| R_0 - R_1 \|_1 \leq \sqrt{14} \| Q^{-1} \|. \]

It follows that
\[ 1 - \text{Tr} \left( \sqrt{Q} R_1 \sqrt{Q} \right) < 2\varepsilon, \]
and therefore
\[ \text{Tr}(X) = \gamma \text{Tr} \left( \sqrt{Q} R_1 \sqrt{Q} \right) \geq (1 - 2\varepsilon) \gamma \]
as required.

Now let us consider the case that the algorithm rejects, and in particular let us focus on the operators \( Y_0, \ldots, Y_{T-1} \) that are computed over the course of the algorithm. As for the case of acceptance, these operators are not viewed as approximations to anything: the fact that these operators exist and cause rejection in the algorithm is enough to conclude that \( x \in A_{\text{no}} \). Let us continue to write
\[ W_{t+1} = \exp \left( -\frac{\varepsilon \gamma \delta}{2} \Phi^s(Y_0 + \cdots + Y_t) \right) \]
and \( \rho_t = W_t / \text{Tr}(W_t) \) for each \( t = 0, \ldots, T - 1 \); but we must keep in mind that the algorithm only computes approximations of these operators. The algorithm must also approximate the spectral decomposition of each \( \Phi(\rho_t) \), where the source of errors in this case comes from both the spectral decomposition computation and the fact that \( \rho_t \) is approximated.
Now, in the error free analysis, the conditions
\[ \text{Tr}(Y_t) \leq \gamma, \quad \langle \rho_t, \Phi^*(Y_t) \rangle = 1, \quad \text{and} \quad \| \delta \Phi^*(Y_t) \| \leq 1 \]
were proved, and these conditions allowed us to conclude that \( Y \) satisfies
\[ \Phi^*(Y) \geq \mathbb{1}_{Y \otimes X} \quad \text{and} \quad \text{Tr}(Y) \leq (1 + \epsilon)\gamma. \]
If we follow precisely the same proof, but with the condition \( \langle \rho_t, \Phi^*(Y_t) \rangle = 1 \) replaced by
\[ \langle \rho_t, \Phi^*(Y_t) \rangle \geq 1 - \alpha \]
for some choice of \( \alpha > 0 \), we once again find immediately that \( \text{Tr}(Y) \leq (1 + \epsilon)\gamma \). This time we have
\[ \lambda_{NM} \left( \Phi^* \left( \frac{Y_0 + \cdots + Y_{T-1}}{T} \right) \right) \geq (1 - \alpha) \exp(-\eta) - \frac{\ln(NM)}{\eta \delta T}, \]
but under the assumption \( \alpha < \eta^2/12 \), say, it follows again that \( \Phi^*(Y) \geq \mathbb{1}_{Y \otimes X} \).
Thus, given that the conditions \( \text{Tr}(Y_t) \leq \gamma \) and \( \| \delta \Phi^*(Y_t) \| \leq 1 \) follow from an inspection of the algorithm as before, it suffices to compute the matrix exponentials and spectral decompositions with sufficient accuracy that \( \langle \rho_t, \Phi^*(Y_t) \rangle > 1 - \eta^2/12 \). This is easily done: as the argument of the matrix exponentials have norm bounded by \( T \), one is able to compute both the matrix exponentials and the spectral decompositions with exponentially greater accuracy in NC than is required.

9 Conclusion

We have proved that \( \text{QIP}(2) \subseteq \text{PSPACE} \) using a semidefinite programming formulation of the maximum acceptance probability of two-message quantum interactive proof systems, along with the multiplicative weights update method for verifying these values. An obvious question remains: can this method be extended, or some other method devised, to prove \( \text{QIP} = \text{PSPACE} \)?

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