Interaction in Quantum Communication Complexity

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February 1, 2008

Abstract

One of the most intriguing facts about communication using quantum states is that these states cannot be used to transmit more classical bits than the number of qubits used, yet there are ways of conveying information with exponentially fewer qubits than possible classically [2, 21]. Moreover, these methods have a very simple structure—they involve little interaction between the communicating parties. We look more closely at the ways in which information encoded in quantum states may be manipulated, and consider the question as to whether every classical protocol may be transformed to a “simpler” quantum protocol of similar efficiency. By a simpler protocol, we mean a protocol that uses fewer message exchanges. We show that for any constant $k$, there is a problem such that its $k+1$ message classical communication complexity is exponentially smaller than its $k$ message quantum communication complexity, thus answering the above question in the negative. Our result builds on two primitives, local transitions in bi-partite states (based on previous work) and average encoding which may be of significance in other applications as well.

1 Introduction

A recurring theme in quantum information processing has been the idea of exploiting the exponential resources afforded by quantum states to encode information in very non-obvious ways. Perhaps the most representative result of this kind is due to Ambainis, Schulman, Ta-Shma, Vazirani and Wigderson [2], which shows that it is possible to deal a random set of $\sqrt{N}$ cards each from a set of $N$ by the exchange of $O(\log N)$ quantum bits between two players. Raz [21] gives a communication problem where the information storage capacity of quantum states is exploited more explicitly. Both are examples of problems for which exponentially fewer quantum bits are required to accomplish a communication task, as compared to classical bits.

The protocols presented by [2, 21] also share the feature that they require minimal interaction between the communicating players. For example, in the first protocol, one player prepares a set of qubits in a certain state and sends half of them across as the message, after which both players measure their qubits...
to obtain the result. On the other hand, efficient quantum protocols for problems such as checking set disjointness (DISJ) seem to require much more interaction: Buhrman, Cleve and Wigderson [4] give an $O(\sqrt{N \log N})$ qubit protocol for DISJ that takes $O(\sqrt{N})$ message exchanges. This represents quadratic savings in communication cost, but also an unbounded increase in the number of messages exchanged (from one message to $\sqrt{N}$), as compared to classical protocols. Can we exploit the features of quantum communication and always reduce interaction while maintaining the same communication cost? In other words, do all efficient quantum protocols have the simple structure shared by those of [2, 21]?

In this paper, we study the effect of interaction on the quantum communication complexity of problems. We show that for any constant $k$, allowing even one more message may lead to an exponential decrease in the communication complexity of a problem, thus answering the above question in the negative. More formally,

**Theorem 1.1** For any constant $k$, there is a problem such that any quantum protocol with only $k$ messages and constant probability of error requires $\Omega(N^{1/(k+1)})$ communication qubits, whereas it can be solved with $k + 1$ messages by a deterministic protocol with $O(\log N)$ bits.

Klauck [11] states a relationship between the bounded message complexity of Pointer Jumping and DISJ. Together with our result, this implies an $\Omega(N^{1/k(k+1)})$ lower bound for $k$ message protocols for DISJ, for any constant $k$.

The role of interaction in classical communication is well-studied, especially in the context of the pointer jumping function [18, 7, 17, 19]. In fact, the problem we study in this paper is the subproblem of Pointer Jumping singled out in [15]. Our analysis has the same gross structure as that in [15] (also explained in [12]), but relies on entirely new ideas from quantum information theory.

In the context of quantum communication, it was observed by Buhrman and de Wolf [5] (based on a lower bound of Nayak [16]) that any one message quantum protocol for DISJ has linear communication complexity. Thus, allowing more interaction leads to a quadratic improvement in communication cost. The lower bound of [16] immediately implies a much stronger separation: it shows that the two message complexity of a problem may be exponentially smaller than its one message complexity (see also [11]). Our result subsumes all these.

Our interest in the role of interaction in quantum communication also springs from the need to better understand the ways in which we can access and manipulate information encoded in quantum states. We develop information-theoretic techniques that expose some of the limitations of quantum communication. More specifically, we present a new primitive in quantum encoding, as suggested by the following theorem.

**Theorem 1.2 (Average encoding theorem)** Let $x \mapsto \sigma_x$ be a quantum encoding mapping $m$ bit strings $x \in \{0,1\}^m$ into mixed states $\sigma_x$. Let $X$ be distributed uniformly over $\{0,1\}^m$, let $Q$ be the encoding of $X$ according to this map, and let $\sigma = \frac{1}{2^m} \sum_x \sigma_x$. Then,

$$\frac{1}{2^m} \sum_x \| \sigma - \sigma_x \|_1 \leq 2 \sqrt{I(Q : X)}.$$

In other words, if an encoding $Q$ is only weakly correlated to a random variable $X$, then the “average encoding” $\sigma$ (corresponding to a random string) is on average a good approximation of any encoded state. Thus, in certain situations, we may dispense with the encoding altogether, and use the single state $\sigma$ instead.

We also use another primitive derived from the work of Lo and Chau [3] and Mayers [14] which combines results of Jozsa [10], and Fuchs and van de Graaf [8]. Consider two bi-partite pure states such that one party sharing the states cannot locally distinguish well between the two states with good probability. Then the other party can locally transform any of the states close to the other.
Theorem 1.3 (Local transition theorem) (based on [13, 14, 10, 8]) Let \( \rho_1, \rho_2 \) be two mixed states with support in a Hilbert space \( \mathcal{H} \), \( \mathcal{K} \) any Hilbert space of dimension at least \( \dim(\mathcal{H}) \), and \( |\phi_i\rangle \) any purifications of \( \rho_i \) in \( \mathcal{H} \otimes \mathcal{K} \). Then, there is a local unitary transformation \( U \) on \( \mathcal{K} \) that maps \( |\phi_2\rangle \rightarrow |\phi'_2\rangle = I \otimes U |\phi_2\rangle \) such that

\[
\| |\phi_1\rangle \langle \phi_1| - |\phi'_2\rangle \langle \phi'_2| \|_1 \leq 2 \| \rho_1 - \rho_2 \|^{\frac{3}{2}}_1.
\]

This may be of significance in cryptographic applications as well.

2 Preliminaries

2.1 The communication complexity model

In the quantum communication complexity model [3], Alice and Bob hold qubits. When the game starts Alice holds a superposition \( |x\rangle \) and Bob holds \( |y\rangle \) (representing the input to the two players), and so the initial joint state is simply \( |x\rangle \otimes |y\rangle \). The two parties then play in turns. Suppose it is Alice’s turn to play. Alice can do an arbitrary unitary transformation on her qubits and then send one or more qubits to Bob. Sending qubits does not change the overall superposition, but rather changes the ownership of the qubits, allowing Bob to apply his next unitary transformation on the newly received qubits. At the end of the protocol, one player makes a measurement and declares that as the result of the protocol.

In general, each player may also (partially) measure her qubits during her turn. However, we assume (by invoking the principle of safe storage [3]) that all such measurements are postponed to the end. We also assume that the two players do not modify the qubits holding the input superposition during the protocol. Neither of these affects the aspect of communication we focus on in this paper.

The complexity of a quantum (or classical) protocol is the number of qubits (respectively, bits) exchanged between the two players. We say a protocol computes a function \( f : \mathcal{X} \times \mathcal{Y} \rightarrow \{0, 1\} \) with \( \epsilon \geq 0 \) error, if for any input \( x \in \mathcal{X}, y \in \mathcal{Y} \) the probability that the two players compute \( f(x, y) \) is at least \( 1 - \epsilon \). \( Q_{\epsilon}(f) \) denotes the complexity of the best quantum protocol that computes \( f \) with at most \( \epsilon \) error.

For a player \( P \in \{ \text{Alice, Bob} \} \), \( Q_{\epsilon}^c(f) \) denotes the complexity of the best quantum protocol that computes \( f \) with at most \( \epsilon \) error with only \( c \) messages, where the first message is sent by \( P \). If the name of the player is omitted from the superscript, either player is allowed to start the protocol.

We say a protocol \( \mathcal{P} \) computes \( f \) with \( \epsilon \) error with respect to a distribution \( \mu \) on \( \mathcal{X} \times \mathcal{Y} \), if

\[
\text{Prob}(x,y)\in\mu, \mathcal{P}(x,y) = f(x,y)) \geq 1 - \epsilon.
\]



\( Q_{\epsilon,c}^\mu(f) \) is the complexity of computing \( f \) with at most \( \epsilon \) error with respect to \( \mu \), with only \( c \) messages where the first message is sent by player \( P \). The following is immediate.

Fact 2.1 For any distribution \( \mu \), number of messages \( c \) and player \( P \), \( Q_{\epsilon,c}^\mu(f) \leq Q_{\epsilon}^c(f) \).

2.2 Classical entropy and mutual information

The Shannon entropy \( S(X) \) of a classical random variable \( X \) quantifies the amount of randomness in it. If \( X \) takes values \( x \) in some finite set with probability \( p_x \), its Shannon entropy is defined as \( S(X) = - \sum_x p_x \log p_x \). The mutual information \( I(X : Y) \) of a pair of random variables \( X, Y \) is defined by \( I(X : Y) = S(X) + S(Y) - S(XY) \). It is a measure of how correlated the two random variables are.
The following are some basic facts about the mutual information function that we use in the paper. For any random variables \(X, Y, Z\),

\[
I(X : Y Z) = I(X : Y) + I(X Y : Z) − I(Y : Z)
\]

(1)

\[
I(X : Y Z) \geq I(X : Y).
\]

(2)

Fano’s inequality states that if \(Y\) can predict another random variable \(X\) with an advantage, then \(X\) and \(Y\) have large mutual information. We use it only in the following simple form.

**Fact 2.2 (Fano’s inequality)** Let \(X\) be a uniformly distributed boolean random variable, and let \(Y\) be a boolean random variable such that \(\text{Prob}(X = Y) \geq \frac{1}{2} + \delta\), where \(\delta \geq 0\). Then \(I(X : Y) \geq 1 − H\left(\frac{1}{2} + \delta\right)\).

For other equivalent definitions and properties of these concepts, we refer the reader to a standard text (such as [6]) on information theory. Finally, we give a simple bound on the deviation of the binary entropy function \(H(p)\) from 1 as \(p\) deviates from 1/2.

**Fact 2.3** For \(\delta \in [-\frac{1}{2}, \frac{1}{2}]\), we have \(H\left(\frac{1}{2} + \delta\right) \leq 1 − \delta^2\).

**Proof:** From the definition of the binary entropy function, we have

\[
H\left(\frac{1}{2} + \delta\right) = 1 - \frac{1}{2}[(1 + 2\delta) \log(1 + 2\delta) + (1 - 2\delta) \log(1 - 2\delta)].
\]

Using the expansion \(\ln(1 + x) = x - \frac{x^2}{2} + \frac{x^3}{3} - \frac{x^4}{4} + \cdots\) for \(|x| < 1\), and simplifying, we get

\[
H\left(\frac{1}{2} + \delta\right) = 1 - (\log e) \left[\left(2 - \frac{2^2}{2 \cdot 2}\right) \delta^2 + \left(\frac{3}{3} - \frac{2^4}{2 \cdot 4}\right) \delta^4 + \left(\frac{5}{5} - \frac{2^6}{2 \cdot 6}\right) \delta^6 + \cdots\right]
\]

\[
\leq 1 - \delta^2,
\]

which is the claimed bound. ■

### 2.3 The density matrix and the trace norm

The quantum mechanical analogue of a random variable is a probability distribution over superpositions, also called a mixed state. Consider the mixed state \(X = \{p_i, |\phi_i\rangle\}\), where the superposition \(|\phi_i\rangle\) is drawn with probability \(p_i\). The density matrix of the mixed state \(X\) is \(\rho_X = \sum_i p_i |\phi_i\rangle \langle \phi_i|\). The following properties of density matrices are immediate from the definition: every density matrix \(\rho\) is Hermitian, i.e., \(\rho = \rho^\dagger\), has unit trace, i.e., \(\text{Tr}(\rho) = \sum_i \rho(i, i) = 1\), and is positive semi-definite, i.e., \(\langle \psi | \rho | \psi \rangle \geq 0\) for all \(|\psi\rangle\). Thus, every density matrix is unitarily diagonalizable and has non-negative real eigenvalues that sum up to 1.

Given a quantum system in a mixed state with density matrix \(\rho\) and a (general) measurement \(\mathcal{O}\) on it, let \(\rho^\mathcal{O}\) denote the classical distribution on the possible results that we get by measuring \(\rho\) according to \(\mathcal{O}\). Suppose that it is some classical distribution \(p_1, \ldots, p_k\) where we get result \(i\) with probability \(p_i\). Given two different mixed states, we can ask how well one can distinguish between the two mixtures, or equivalently, how different the distributions resulting from a measurement may be. To quantify this, we consider the \(\ell_1\) metric: if \(p = (p_1, \ldots, p_k)\) and \(q = (q_1, \ldots, q_k)\) are two probability distributions over \(\{1, \ldots, k\}\), then the \(\ell_1\) distance between them is \(\|p - q\|_1 = \sum_i |p_i - q_i|\). A fundamental theorem about distinguishing density matrices (see [3]) tells us:
Theorem 2.4 Let $\rho_1, \rho_2$ be two density matrices on the same space $\mathcal{H}$. Then for any (general) measurement $O$

$$\left\| \rho_1^O - \rho_2^O \right\|_1 \leq \text{Tr} A^\dagger A,$$

where $A = \rho_1 - \rho_2$. Furthermore, the bound is tight, and the orthogonal measurement $O$ that projects a state on the eigenvectors of $\rho_1 - \rho_2$ achieves this bound.

Theorem 2.4 shows that the density matrix captures all the accessible information that a quantum state contains. If two different mixtures have the same density matrix (which is indeed possible) then even though they are two distinct distributions, they are physically, and thus from a computational point of view, indistinguishable. As the behavior of a mixed state is completely characterized by its density matrix we often identify a mixed state with its density matrix.

The quantity $\text{Tr} A^\dagger A$ is of independent interest. (Note that this is compact notation for the sum of the (magnitudes of the) singular values of $A$.) If we define $\| A \|_t = \text{Tr} A^\dagger A$ then $\| \cdot \|_t$ defines a norm (the trace norm), and has some additional properties such as $\| A \otimes B \|_t = \| A \|_t \cdot \| B \|_t$, $\| A \|_t = 1$ for any density matrix $A$ and $\| AB \|_t, \| BA \|_t \leq \| A \|_t \cdot \| B \|_t$. (See [1] for more details.) We single out the following fact for later use.

Fact 2.5 If $|\phi_1\rangle, |\phi_2\rangle$ are two pure states, and $\rho_i$ is the density matrix of $|\phi_i\rangle$, then

$$\| \rho_1 - \rho_2 \|_t = 2\sqrt{1 - |\langle \phi_1 | \phi_2 \rangle|^2}.$$ 

2.4 The fidelity measure

A useful alternative to the trace metric as a measure of closeness of density matrices is fidelity, which is defined in terms of the pure states that can give rise to those density matrices. A purification of a mixed state $\rho$ with support in a Hilbert space $\mathcal{H}$ is any pure state $|\phi\rangle$ in an extended Hilbert space $\mathcal{H} \otimes \mathcal{K}$ such that $\text{Tr}_\mathcal{K} |\phi\rangle \langle \phi| = \rho$. Given two density matrices $\rho_1, \rho_2$ on the same Hilbert space $\mathcal{H}$, their fidelity is defined as

$$F(\rho_1, \rho_2) = \sup |\langle \phi_1 | \phi_2 \rangle|^2,$$

where the supremum is taken over all purifications $|\phi_1\rangle$ of $\rho_i$ in the same Hilbert space [10]. We state a few properties of this measure: $0 \leq F(\rho_1, \rho_2) \leq 1$, $F(\rho_1, \rho_2) = 1 \iff \rho_1 = \rho_2$ and if $\rho_1 = |\phi_1\rangle \langle \phi_1|$, then we have $F(\rho_1, \rho_2) = |\langle \phi_1 | \phi_2 \rangle|^2$. Jozsa [10] proved that the optimum is always achieved when finite dimensional density matrices are considered.

Theorem 2.6 (Jozsa) Let $\rho_1, \rho_2$ be any two mixed states with support in a finite dimensional Hilbert space $\mathcal{H}, \mathcal{K}$ a Hilbert space of dimension at least $\dim(\mathcal{H})$, and $|\phi_1\rangle$ any purification of $\rho_1$ in $\mathcal{H} \otimes \mathcal{K}$. Then there exists a purification $|\phi_2\rangle \in \mathcal{H} \otimes \mathcal{K}$ of $\rho_2$ such that $|\langle \phi_1 | \phi_2 \rangle|^2 = F(\rho_1, \rho_2)$.

Jozsa [10] also gave a simple proof (again for the finite dimensional case) of the following remarkable equivalence first established by Uhlmann [22].

$$F(\rho_1, \rho_2) = \left[ \text{Tr} \left( \sqrt{\rho_1 \rho_2 \rho_1^\dagger} \right)^\frac{1}{2} \right]^2 = \| \sqrt{\rho_1} \sqrt{\rho_2} \|^2_t.$$

Using this equivalence, Fuchs and van de Graaf [8] show that the fidelity and the trace measures of distance between density matrices are closely related. They prove:

Theorem 2.7 (Fuchs, van de Graaf) For any two mixed states $\rho_1, \rho_2$,

$$1 - F(\rho_1, \rho_2) \leq \frac{1}{2} \| \rho_1 - \rho_2 \|_t \leq \sqrt{1 - F(\rho_1, \rho_2)}.$$
2.5 Von Neumann entropy and quantum mutual information

As mentioned earlier, the eigenvalues of a density matrix are all real, non-negative and sum up to one. Thus, they induce a probability distribution on the corresponding eigenvectors. Since the eigenvectors are all orthogonal, this is essentially a classical distribution. Every mixed state with the same density matrix is physically equivalent to such a canonical classical distribution. It is thus natural to define the entropy of a mixed state as the Shannon entropy of this distribution. Formally, the von Neumann entropy $S(\rho)$ of a density matrix $\rho$ is defined as $S(\rho) = -\sum \lambda_i \log \lambda_i$, where $\{\lambda_i\}$ is the multi-set of all the eigenvalues of $\rho$. More compactly, $S(\rho) = -\text{Tr} \rho \log \rho$.

Not all properties of classical Shannon entropy carry over to the quantum case. For example it is quite possible that $S(XY) < S(X)$ as can be seen by considering the pure state $\frac{1}{\sqrt{2}}(|0\rangle_X |0\rangle_Y + |1\rangle_X |1\rangle_Y)$. Nonetheless, some of the classical properties do carry over, e.g., $S(X) \geq 0$, $S(X)$ is concave and $S(XY) \leq S(X) + S(Y)$. A property of interest to us is the following, which also generalizes a classical assertion.

**Fact 2.8** Suppose a quantum system $A$ is in mixed state $\{p_i, |i\rangle\}$, where $\{|i\rangle\}$ are orthogonal, and $\sigma_i$ are density matrices for another system $B$, then $S(\sum_i p_i |i\rangle \langle i| \otimes \sigma_i) = H(A) + \sum_i p_i S(\sigma_i)$.

The density matrix corresponding to a mixed state with superpositions drawn from a Hilbert space $\mathcal{H}$ is said to have support in $\mathcal{H}$. A density matrix with support in a Hilbert space of dimension $d$, has $d$ eigenvalues, hence the entropy of any such distribution is at most $\log d$. A pure-state has zero entropy. Measuring a pure-state may result in a non-trivial mixture and positive entropy. In general, orthogonal measurements increase the entropy. For a comprehensive introduction to this concept and its properties see, for instance, [2].

We define the “mutual information” $I(X : Y)$ of two disjoint systems $X, Y$ in analogy with classical mutual information: $I(X : Y) = S(X) + S(Y) - S(XY)$, where $XY$ is density matrix of the system that includes the qubits of both systems. Again, not all properties of classical mutual information carry over to the quantum case. For example, it is not true in general that $I(X : Y) \leq S(X)$. Nonetheless, some of the intuition we have about mutual information still applies. Equation (1) still holds, as follows immediately from the definition. Equation (2) also continues to be true, but its proof is much more involved. It is in fact equivalent to the strong sub-additivity property of von Neumann entropy. An important consequence of this property is that local measurements can only decrease the amount of mutual information. A special case of this is the classic Holevo theorem [3] from quantum information theory, which bounds the amount of information we can extract from a quantum encoding of classical bits.

**Theorem 2.9 (Holevo)** Let $x \mapsto \sigma_x$ be any quantum encoding of bit strings into density matrices, let $X$ be a random variable with a distribution given by $\text{Prob}(X = x) = p_x$, let $Q$ be the quantum encoding of $X$ according to this map, and let $\sigma = \sum_x p_x \sigma_x$. If $Y$ is any random variable obtained by performing a measurement on the encoding, then

$$I(X : Y) \leq I(X : Q) = S(\sigma) - \sum_x p_x S(\sigma_x).$$

In analogy with classical conditional entropy, we define $S(Y | X) = \sum_x p_x S(\sigma_x | X)$, where $X$ is a classical random variable and $Y$ is a quantum encoding of it given by $x \mapsto \sigma_x$. We similarly define conditional von Neumann entropy and mutual information with respect to a classical event. Thus, for example, $I(X : Y) = S(Y) - S(Y | X)$. 


3 The average encoding theorem

The average encoding theorem asserts that if a quantum encoding has little correlation with the encoded classical information then the encoded states are essentially indistinguishable. In particular, they are all “close” to the average encoding. This theorem formalizes a very intuitive idea and might seem to be immediate from Holevo’s theorem. However, there is a subtle difference: in Holevo’s theorem one is interested in a single measurement that simultaneously distinguishes all the states, whereas in our case we are interested in the pairwise distinguishability of the encoded states. We first prove:

**Theorem 3.1** Let \( x \mapsto \sigma_x \) be a quantum encoding mapping \( m \) bit strings \( x \in \{0, 1\}^m \) into mixed states \( \sigma_x \). Let \( X \) be distributed uniformly over \( \{0, 1\}^m \) and let \( Q \) be the encoding of \( X \) according to this map. Denote \( \Delta = \frac{1}{2m} \sum_{x_1, x_2 \in \{0, 1\}^m} \| \sigma_{x_1} - \sigma_{x_2} \|_1 \). Then \( I(X : Q) \geq 1 - H\left(\frac{1 + \Delta}{2}\right) \).

**Proof:** We start with the special case of \( m = 1 \). By Theorem 2.4, there is a measurement \( O \) on \( Q \) that realizes the trace norm distance \( t = \| \sigma_0 - \sigma_1 \|_t \) between \( \sigma_0 \) and \( \sigma_1 \). Using Bayes’ strategy (see, for example, [8]), the resulting distributions can be distinguished with probability \( \frac{1}{2} + \frac{t}{4} \). Let \( Y \) denote the classical random variable holding the result of this entire procedure. We have \( \text{Prob}(Y = X) = \frac{1}{2} + \frac{t}{4} \).

Thus, by Fano’s Inequality,

\[
I(X : Y) \geq 1 - H\left(\frac{1}{2} + \frac{t}{4}\right)
\]

We complete the proof for \( m = 1 \) by noticing that measurements can only reduce the entropy, hence \( I(X : Q) \geq I(X : Y) \), and that \( \Delta = \frac{t}{2} \).

To prove the theorem for general \( m \) we reduce it to the \( m = 1 \) case. We do this by partitioning the set of strings into pairs with “easily” distinguishable encoding.

**Lemma 3.2** There is a set of \( 2^m/2 \) disjoint pairs \( \{x_{2i-1}, x_{2i}\} \) which together cover \( \{0, 1\}^m \) such that

\[
\frac{2}{2m} \sum_i \| \sigma_{x_{2i-1}} - \sigma_{x_{2i}} \|_1 \geq \Delta.
\]

**Proof:** The expectation of the LHS over a random pairing is \( \frac{2^m}{2m} \Delta \); hence there is a pairing that achieves this \( \Delta \).

We now fix this pairing. Let \( Z_i \) denote the set of elements in the \( i \)’th pair, i.e., \( Z_i = \{x_{2i-1}, x_{2i}\} \) and \( \Delta_i = \| \sigma_{x_{2i-1}} - \sigma_{x_{2i}} \|_1 \). We know that \( \frac{2^m}{2m} \sum \Delta_i \geq \Delta \). Let us also denote \( f(\delta) = 1 - H(\frac{1 + \delta}{2}) \). From the base case \( m = 1 \), we know that for any \( i = 1, \ldots, 2^m/2 \), \( I(X : Q \mid X \in Z_i) \geq f(\Delta_i) \). Thus we get:

\[
S(Q \mid X \in Z_i) - \frac{1}{2}[S(\sigma_{x_{2i}}) - S(\sigma_{x_{2i+1}})] \geq f(\Delta_i).
\]

Averaging all the \( 2^m/2 \) equations yields:

\[
\frac{2}{2m} \sum_i S(Q \mid X \in Z_i) - \frac{1}{2m} \sum_x S(\sigma_x) \geq \frac{2}{2m} \sum_i f(\Delta_i)
\]

By the concavity of the entropy function, \( S(Q) \geq \frac{2}{2m} \sum_i S(Q \mid X \in Z_i) \), and by definition \( \frac{1}{2m} \sum_x S(\sigma_x) = S(Q \mid X) \). Therefore,

\[
I(X : Q) = S(Q) - S(Q \mid X) \geq \frac{2}{2m} \sum_i f(\Delta_i).
\]
Thus, $\Delta'\leq 2$ via Theorem 2.6, since the trace distance is zero. Also, $f(\Delta)$ is monotone increasing for $0 \leq \delta \leq \frac{1}{2}$, so $f(\frac{1}{2m}\sum_{i=1}^{m} \Delta_i) \geq f(\Delta)$. Together this yields $I(X:Q) \geq f(\Delta)$, as required. \hfill \Box

Now, we can easily deduce Theorem 1.2.

**Proof of Theorem 1.2:** Let $\Delta' = \frac{1}{2m}\sum_{x_1} \| \sigma_{x_1} - \sigma \|_t$. We have:

$$\Delta' = \frac{1}{2m}\sum_{x_1} \| \sigma_{x_1} - \sigma \|_t = \frac{1}{2m}\sum_{x_1} \left\| \frac{1}{2m} \sum_{x_2} (\sigma_{x_1} - \sigma_{x_2}) \right\|_t \leq \frac{1}{2^m} \sum_{x_1,x_2} \| \sigma_{x_1} - \sigma_{x_2} \|_t \leq \Delta$$

By Theorem 2.7, $I(X:Q) \geq 1 - H(\frac{1+\Delta}{2})$, and by Fact 2.3 we have $1 - H(\frac{1+\Delta}{2}) \geq 1 - (1 - \frac{\Delta}{2})^2 = \frac{\Delta^2}{4}$. Thus, $\Delta' \leq \Delta \leq 2\sqrt{I(X:Q)}$. \hfill \Box

## 4 Local transition between bipartite states

Lo and Chau [13] and Mayers [14] proved:

**Theorem 4.1 (Lo and Chau; Mayers)** Suppose $|\phi_1\rangle$ and $|\phi_2\rangle$ are two pure states in the Hilbert space $H \otimes K$, such that $\text{Tr}_K |\phi_2\rangle \langle \phi_2| = \text{Tr}_K |\phi_1\rangle \langle \phi_1|$, i.e., the reduced density matrix of $|\phi_2\rangle$ to $H$ is the same as the reduced density matrix of $|\phi_1\rangle$ to $H$. Then, there is a local unitary transformation $U$ on $K$ such that $I \otimes U |\phi_2\rangle = |\phi_1\rangle$.

The theorem follows by examining the Schmidt decomposition [20] of the two states.

A natural generalization of this is to the case where the reduced density matrices are close to each other but not quite the same, which is what appears in Theorem 1.3. Lo and Chau [13] and Mayers [14] considered this case as well. Theorem 1.3 follows from their work by using the newer results of [8] stated in Theorem 2.5.

**Proof of Theorem 1.3:** By Theorem 2.4, there exists a purification $|\phi'_2\rangle \in H \otimes K$ of $\rho_2$ such that $|\langle \phi_1| \phi'_2\rangle|^2 = F(\rho_1, \rho_2)$. Since $|\phi_2\rangle$ and $|\phi'_2\rangle$ have the same reduced density matrix in $H$, by Theorem 4.1 there is a (local) unitary transformation $U$ on $K$ such that $I \otimes U |\phi_2\rangle = |\phi'_2\rangle$. Moreover, by Fact 2.3 we have

$$\| |\phi_1\rangle \langle \phi_1| - |\phi_2\rangle \langle \phi_2| \|_t = 2\sqrt{1 - |\langle \phi_1| \phi_2\rangle|^2} = 2\sqrt{1 - F(\rho_1, \rho_2)}.$$

By Theorem 2.7, $\sqrt{F(\rho_1, \rho_2)} \geq 1 - \frac{1}{2} \rho_1 - \rho_2 \|_t$, so

$$1 - F(\rho_1, \rho_2) \leq 1 - \left(1 - \frac{1}{2} \rho_1 - \rho_2 \|_t \right)^2 \leq \rho_1 - \rho_2 \|_t.$$

This, when combined with the earlier bound on the trace distance between $|\phi_1\rangle$, $|\phi'_2\rangle$ gives us the required result. \hfill \Box

## 5 The role of interaction in quantum communication

In this section, we prove that allowing more interaction between two players in a quantum communication game can substantially reduce the amount of communication required. We first define a communication problem and state our results formally (giving an overview of the proof), and then give the details of the proofs.
5.1 The communication problem and its complexity

In this section, we give the main components of the proof of Theorem 5.1. We define a sequence of problems $S_0, S_1, \ldots, S_k, \ldots$ by induction. The problem $S_1$ is the index function, i.e., Alice has a $n$-bit string $x \in X_1 = \{0,1\}^n$, Bob has an index $i \in Y_1 = [n]$ and the desired output is $S_1(x, i) = x_i$. Suppose we have already defined the function $S_{k-1} : X_{k-1} \times Y_{k-1} \rightarrow \{0,1\}$. In the problem $S_k$, Alice has as input her part of $n$ independent instances of $S_{k-1}$, i.e., $x \in X_{k-1}^n$. Bob has his share of $n$ independent instances of $S_{k-1}$, i.e., $y \in Y_{k-1}^n$, and in addition, there is an extra input $a \in [n]$ which is given to Alice if $k$ is even and to Bob if $k$ is odd. The output we seek is the solution to the $a$th instance of $S_{k-1}$. In other words, $S_k(x_1, \ldots, x_n, a, y_1, \ldots, y_n) = S_{k-1}(x_a, y_a)$.

Note that the input size to the problem $S_k$ is $N = \Theta(n^k)$.

In fact, we prove a stronger intermediate claim. Let $P_1$ be Alice, and for $k \geq 2$, let $P_k$ denote the player that holds the index $a$ in an instance of $S_k$ ($a$ indicates which of the $n$ instances of $S_{k-1}$ to solve). Let $P_k$ denote the other player. We refer to $P_k$ as the “wrong” player to start a protocol for $S_k$. The stronger claim is that any $k$-message protocol for $S_k$ in which the wrong player starts is exponentially inefficient as compared to the $\log N$ protocol described above.

**Theorem 5.1** For all constant $k \geq 1$, $0 \leq \epsilon < \frac{1}{2}$, $Q_{U,\epsilon}^k(S_{k+1}) \geq \Omega \left( (N^{1/(k+1)}) \right)$.

In fact, there is a classical $k$-message protocol in which the wrong player starts with complexity $O(n)$, so our lower bound is optimal.

**Theorem 5.1** now follows directly.

**Proof of Theorem 5.1**: It is enough to show the lower bound for the two cases when the protocol starts either with $P_{k+1}$ or with the other player.

Let $P_{k+1}$ be the player to start. Note that if we set $a$ to a fixed value, say 1, then we get an instance of $S_k$. So $Q_{U,\epsilon}^{k,P_{k+1}}(S_k) \geq Q_{U,\epsilon}^{k,P_{k+1}}(S_{k+1})$. But $P_{k+1} = P_k$, so the bound of Theorem 5.1 applies.

Let player $P_{k+1}$ be the one to start. Then, observe that if we allow one more message (i.e., $k + 1$ messages in all), the complexity of the problem only decreases: $Q_{U,\epsilon}^{k+1,P_{k+1}}(S_{k+1}) \leq Q_{U,\epsilon}^{k,P_{k+1}}(S_{k+1})$. So we again get the same bound from Theorem 5.1.

We prove Theorem 5.1 by induction. First, we show that the index function is hard to solve with one message if the wrong player starts. This essentially follows from the lower bound for random access codes in [10]. The only difference is that we seek a lower bound for a protocol that has low error probability on average rather than in the worst case, so we need a refinement of the original argument. We give this in the next section.

**Lemma 5.3** For any $0 \leq \epsilon \leq 1$, $Q_{U,\epsilon}^{1,A}(S_1) \geq (1 - H(\epsilon))n$.

Next, we show that if we can solve $S_k$ with $k$ messages with the wrong player starting, then we can also solve $S_{k-1}$ with only $k - 1$ messages of almost the same total length, again with the wrong player starting, at the cost of a slight increase in the average probability of error.
Lemma 5.4 For all $k \geq 2$, $0 \leq \epsilon < \frac{1}{2}$, $Q_{U,\epsilon}^{k-1,\bar{P}_{k-1}}(S_{k-1}) \leq \ell + \log n$, where $\ell = Q_{U,\epsilon}^k(S_k)$, and $\epsilon' = \epsilon + 4(\ell/n)^{1/4}$.

We defer the proof of this lemma to a later section, but show how it implies Theorem 5.2 above.

Proof of Theorem 5.2: We prove the theorem by induction on $k$. The case $k = 1$ is handled by Lemma 5.3. Suppose the theorem holds for $k - 1$. We prove by contradiction that it holds for $k$ as well.

If $Q_{U,\epsilon}^k(S_k) = o(n)$, then by Lemma 5.4 there is a $k - 1$ message protocol for $S_{k-1}$ with the wrong player starting, with error $\epsilon' = \epsilon + o(1) < \frac{1}{2}$, and with the same communication complexity $o(n)$. This contradicts the induction hypothesis. ■

5.2 Hardness of the index function

We now prove the average case hardness of the index function.

Proof of Lemma 5.3: Let $Q$ denote the message sent by Alice. For a prefix $y \in \{0,1\}^j$ of length $i \geq 0$, let $Q_y$ be the encoding which is prepared by first fixing $x_1 = y_1, \ldots, x_i = y_i$ and then choosing $x_{i+1}, \ldots, x_m$ at random and sending the state $\sigma_x$. Its density matrix is given by

$$
\sigma_y = \frac{1}{2m-i} \sum_{z \in \{0,1\}^{m-i}} \sigma_{yz}.
$$

On the one hand, $I(Q : X) \leq \ell$, the number of qubits in $Q$. On the other hand, for $y \in \{0,1\}^j$, let $\epsilon_y$ be the error probability when $x_i = y_i$, $i \leq j$, and the index $i = j + 1$. Note that $\epsilon = \frac{1}{n} \sum_{j=0}^{n-1} \frac{1}{2^j} \sum_{y \in \{0,1\}^j} \epsilon_y$.

Moreover, we have $I(Q_y : X_{j+1}) \geq 1 - H(\epsilon_y)$, since Bob has a measurement that predicts $X_{j+1}$ with probability $1 - \epsilon_y$ given $Q_y$. We now claim that

Lemma 5.5 $\frac{1}{2^j} \sum_y \sum_{i=0}^{m-1} I(Q_{x_1 \cdots x_i} : X_{i+1}) \leq I(Q : X)$.

By this lemma,

$$
I(Q : X) \geq \sum_{i=0}^{n-1} \frac{1}{2^i} \sum_{y \in \{0,1\}^j} I(Q_y : X_{j+1}) \geq (1 - H(\epsilon))n,
$$

using the concavity of the entropy function. ■

Proof of Lemma 5.3: By the definition of mutual information, and using Fact 2.8,

$$
I(QX_1 \cdots X_i : X_{i+1}) = S(QX_1 \cdots X_i) + S(X_{i+1}) - S(QX_1 \cdots X_{i+1})
$$

$$
= [i + \frac{1}{2^i} \sum_{y \in \{0,1\}^j} S(\sigma_y)] + [1] - [(i + 1) + \frac{1}{2^{i+1}} \sum_{y \in \{0,1\}^{i+1}} S(\sigma_y)]
$$

$$
= \frac{1}{2^i} \sum_{y \in \{0,1\}^i} [S(\sigma_y) - \frac{1}{2} (S(\sigma_y) + S(\sigma_{y0}) + S(\sigma_{y1}))]
$$

$$
= \frac{1}{2^i} \sum_{y \in \{0,1\}^i} I(Q_y : X_{i+1}).
$$

Moreover, from Properties (1) and (2),

$$
I(Q : X) \geq \sum_{i=0}^{m-1} I(Q, X_1, \ldots, X_i : X_{i+1})
$$
chosen uniformly at random from $Y$. Resulting distribution

**Lemma 5.7** There is a protocol $y$ does not carry any information about $A$. This gives a protocol for solving $S$. This only affects the average probability of error. Moreover, in $P'$ the first message does not carry any information about $y_j$ and is therefore completely independent of it. Intuitively this means that Alice does not need to get that message at all, or equivalently that she can recreate it herself. This gives a protocol for solving $S_{k-1}(x_j, y_j)$ with $k-1$ messages and with Alice starting.

**Lemma 5.6** There is a protocol $P'$ which solves $S_k$ with respect to the distribution $U_{a=j}$ with error $\delta_j = \epsilon_j + 4\mu_j^{1/4}$ error, $\ell$ message qubits and $k$ rounds starting with Bob, such that $I(M : Y_j) = 0$.

The protocol $P'$ is obtained by slightly modifying the first message in protocol $P$ so that it is completely independent of $Y_j$. This only affects the average probability of error. Moreover, in $P'$ the first message does not carry any information about $y_j$ and is therefore completely independent of it. Intuitively this means that Alice does not need to get that message at all, or equivalently that she can recreate it herself. This gives a protocol for solving $S_{k-1}(x_j, y_j)$ with $k-1$ messages and with Alice starting.

**Lemma 5.7** There is a protocol $P''$ that solves $S_{k-1}$ with respect to $U$ with $\epsilon'$ error, $\ell + \log n$ message qubits and $k-1$ messages starting with Alice.

Together we get $Q_{U;A}^{k-1,A}(S_{k-1}) \leq \ell + \log n$ as claimed. 

\[
= \sum_{i=0}^{m-1} \frac{1}{2^i} \sum_{y \in \{0,1\}^i} I(Q_y : X_{i+1}) \\
= \sum_{i=0}^{m-1} \frac{1}{2^m} \sum_{y \in \{0,1\}^m} I(Q_{y_1 \ldots y_i} : X_{i+1}),
\]

which proves the claim. 

### 5.3 The reduction step

In this section, we show how an efficient protocol for $S_k$ gives rise to an efficient protocol for $S_{k-1}$. The gross structure of the argument is the same as in [13, 12]. However, we use entirely new techniques from quantum information theory, as developed in Section 3 and 4 and also get better bounds in the process.

**Proof of Lemma 5.4:** For concreteness, we assume that $k$ is even, so that $P_k$ is Bob. Let $P$ be a protocol that solves $S_k$ with respect to $U$ with $\ell$ message qubits, error $\epsilon$, and $k$ messages starting with Bob. We would like to concentrate on inputs where $x_i$ is chosen uniformly at random from $X_k$ when run on the mixed state $U_\delta$. This only affects the average probability of error. Moreover, in $P'$ the first message does not carry any information about $y_j$ and is therefore completely independent of it. Intuitively this means that Alice does not need to get that message at all, or equivalently that she can recreate it herself. This gives a protocol for solving $S_{k-1}(x_j, y_j)$ with $k-1$ messages and with Alice starting.

**Proof of Lemma 5.6:** Let $\epsilon_j$ denote the error of $P$ with respect to the distribution $U_{a=j}$. Note that $\frac{1}{n} \sum_i \epsilon_i \leq \epsilon$, since having the $Y_i$ in a uniform superposition over all possible inputs has the same effect on the result of the protocol as having it randomly distributed over the inputs (recall that we require that the input registers are not changed during a quantum protocol). Let $\mu_j$ be the mutual information $I(M : Y_j)$ in the protocol $P$ when run on the mixed state $U_{a=j}$ with $y_j$ being chosen randomly.

**Lemma 5.6** There is a protocol $P'$ which solves $S_k$ with respect to the distribution $U_{a=j}$ with error $\delta_j = \epsilon_j + 4\mu_j^{1/4}$ error, $\ell$ message qubits and $k$ rounds starting with Bob, such that $I(M : Y_j) = 0$.
5.4 Proof of Lemmas 5.6 and 5.7

Proof of Lemma 5.6: First consider the case when \( Y_j \) is fixed to some \( z \), but the rest of the inputs are as in \( U_{a=j} \). In protocol \( P \) Bob applies a unitary transformation \( V \) on his qubits and computes \( \langle \phi(z) \rangle = V |0, Y_1, \ldots, Y_n \rangle \) in register \( M \) (for the message) and \( B \) (for Bob’s ancilla and input). In \( P' \) the message computation is slightly different. Instead of computing \( \langle \phi(z) \rangle \), Bob computes \( \langle \phi' \rangle = V |0, Y_1, \ldots, Y_{j-1}\rangle |\psi \rangle |Y_{j+1}, \ldots, Y_n\rangle \), where \( |\psi \rangle \) is the uniform superposition over \( \mathcal{Y}_{k-1} \). Clearly, in \( P' \) the state \( \langle \phi' \rangle \) and hence the message \( M \) does not depend on \( y_j = z \), hence \( I(M : Y_j) = 0 \) when \( y_j \) is chosen randomly.

Let us denote by \( \rho_M(z) \) the reduced density matrix of the message register \( M \) in \( P \) when the input is drawn according to \( U_{a=j} \) and \( y_j = z \), and let the corresponding density matrix for \( P' \) be \( \rho_M \). Clearly, \( \rho_M = \frac{1}{|\mathcal{Y}_{k-1}|} \sum_{z \in \mathcal{Y}_{k-1}} \rho_M(z) \). Let \( \tau_z = \| \rho_M - \rho_M(z) \|_t \). By Theorem 1.2 we know that \( \tau_z \leq 2 \sqrt{\mu_j} \).

Protocol \( P' \) generates the pure state \( \langle \phi' \rangle \), while the desired pure state is \( \langle \phi(z) \rangle \). Bob, who knows \( y_j = z \) knows both \( \langle \phi(z) \rangle \) and \( \langle \phi' \rangle \). By Theorem 1.2 there is a local unitary transformation \( T_z \) acting on register \( B \) alone, such that

\[
\| T_z \phi' \langle T_z \phi' - \langle \phi(z) \rangle \langle \phi(z) \|_t \leq 2 \sqrt{\tau_z}.
\]

The next step in protocol \( P' \) is that Bob applies the transformation \( T_z \) to his register \( B \). After that, protocol \( P' \) proceeds exactly as in \( P \). Therefore, for a given \( z \), the probability that \( P \) and \( P' \) disagree on the result is at most \( 2 \sqrt{\tau_z} \), and the error probability of \( P' \) on \( U_{a=j} \) is at most

\[
\delta_j = \epsilon_j + 2 \epsilon z \sqrt{\tau_z} \leq \epsilon_j + 2 \sqrt{E_z \tau_z} \leq \epsilon_j + 4 \frac{1}{\sqrt{\mu_j}},
\]

where the second step follows from Jensen’s inequality.

Proof of Lemma 5.7: Protocol \( P'' \) solves an instance of \( S_{k-1} \). Alice is given an input \( \hat{x} \in_R \mathcal{X}_{k-1} \) and Bob is given an input \( \hat{y} \in_R \mathcal{Y}_{k-1} \). The protocol proceeds as follows. Alice and Bob first reduce the problem to an \( S_k \) instance taken from the distribution \( U_{a=j} \) for a random \( j \). To do that, Alice picks \( j \in [n] \) at random, sets \( a = j \) and sends it to Bob; Alice sets \( x_j = \hat{x} \) and Bob sets \( y_j = \hat{y} \); Alice picks \( x_i \in_R \mathcal{X}_{k-1} \) for \( i \neq j \); and Bob initializes each register \( Y_i \) for \( i \neq j \) with \( \sum_{z \in \mathcal{Y}_{k-1}} |z \rangle \) (normalized).

Notice that if Alice and Bob run the protocol \( P' \) over this input, then they get the answer \( S_{k-1}(x, y) \) with probability of error at most

\[
\epsilon' = \frac{1}{n} \sum_{i=1}^n \delta_i \leq \frac{1}{n} \sum_{i=1}^n \epsilon_i + 4 \frac{1}{n} \sum_{i=1}^n \mu_i \frac{1}{\sqrt{\mu_j}} \leq \epsilon + 4 \left[ \frac{1}{n} \sum_{i=1}^n \mu_i \right] \frac{1}{\sqrt{n}}.
\]

We claim that

Claim 5.8 \( \sum \mu_i \leq \ell_1 \), where \( \ell_1 \) is the length of the message \( M \).

Hence \( \epsilon' \leq \epsilon + 4(\ell/n)^{1/4} \).

Alice and Bob do not run the protocol \( P' \) itself, but a modification of it in which Alice sends the first message instead of Bob, thus reducing the number of rounds to \( k - 1 \).

Let \( \rho_M \) be the reduced density matrix of register \( M \) holding the first message that Bob sends to Alice in \( P' \), for the input given above. By Lemma 5.6 we know that \( \rho_M \) does not depend on \( y_j = \hat{y} \). So \( \rho_M \) is known in advance to Alice. Alice starts the protocol \( P'' \) by purifying \( \rho_M \). More specifically, let \( \{|e_i\} \) be an eigenvector basis for \( \rho_M \) with real and positive eigenvalues \( \lambda_i \). Alice constructs the
superposition \( \sum \sqrt{\lambda_i} |e_i, i \rangle \) over two registers \( M \) (containing the eigenvectors) and \( B \) (containing the index \( i \)), and sends register \( B \) to Bob. The state of the system after this message in \( P'' \) is

\[
|\xi\rangle = |x_1, \ldots, x_n\rangle_A \otimes \sum_i \sqrt{\lambda_i} |e_i\rangle_M |i\rangle_B
\]

whereas in \( P' \) it is

\[
|\chi(y)\rangle = |x_1, \ldots, x_n\rangle_A \otimes |T_y \phi'\rangle_{MB}.
\]

The reduced density matrix of \( |\xi\rangle \) to registers \( AM \) is the same as the reduced density matrix of \( |\chi(y)\rangle \) to registers \( AM \). By Theorem 4.1, Bob has a local unitary transformation \( V_y \) (operating on his register \( B \)) that transforms \( |\xi\rangle \) to \( |\chi(y)\rangle \). Bob applies \( V_y \), and Alice and Bob then simulate the rest of the protocol \( P' \). From this stage on, the runs of the protocols \( P' \) and \( P'' \) are identical, have the same communication complexity and success probability.

**Proof of Claim 5.8:** Note that \( \mu_j \) is the same as the mutual information \( I(M : Y_j) \) when \( P \) is run on the uniform distribution on \( A_{k-1}^n \times Y_{k-1}^n \). So we prove the claim for the latter.

For any \( i \), \( I(Y_i : Y_1 \cdots Y_{i-1} Y_{i+1} \cdots Y_n) = 0 \). Therefore by Properties (1) and (2) (cf. Section 2) we have

\[
I(M : Y_1 \cdots Y_n) \geq \sum_{i=1}^{n} I(MY_1 \cdots Y_{i-1} : Y_i) \geq \sum_{i=1}^{n} I(M : Y_i) = \sum_{i} \mu_i
\]

As the first message \( M \) contains only \( \ell_1 \) qubits, we have \( \sum_i \mu_i \leq I(M : Y_1 \cdots Y_n) \leq \ell_1 \).

**Acknowledgements**

We thank Jaikumar Radhakrishnan and Venkatesh Srinivasan for their input on the classical communication complexity of the pointer jumping and the subproblem we study in this paper, and Dorit Aharonov for helpful comments.

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