Estimating distinguishability measures on quantum computers

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The performance of a quantum information processing protocol is ultimately judged by distinguishability measures that quantify how distinguishable the actual result of the protocol is from the ideal case. The most prominent distinguishability measures are those based on the fidelity and trace distance, due to their physical interpretations. In this paper, we propose and review several algorithms for estimating distinguishability measures based on trace distance and fidelity. The algorithms can be used for distinguishing quantum states, channels, and strategies (the last also known in the literature as “quantum combs”). The fidelity-based algorithms offer novel physical interpretations of these distinguishability measures in terms of the maximum probability with which a single prover (or competing provers) can convince a verifier to accept the outcome of an associated computation. We simulate many of these algorithms by using a variational approach with parameterized quantum circuits. We find that the simulations converge well in both the noiseless and noisy scenarios, for all examples considered. Furthermore, the noisy simulations exhibit a parameter noise resilience. Finally, we establish a strong relationship between various quantum computational complexity classes and distance estimation problems.

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In quantum information processing, it is essential to quantify the performance of protocols by using distinguishability measures. It is typically the case that there is an ideal state to prepare or an ideal channel to simulate, but in practice, we can only realize approximations, due to experimental error. Two commonly employed distinguishability measures for states are the trace distance [Hel67, Hel69] and the fidelity [Uhl76]. The former has an operational interpretation as the distinguishing advantage in the optimal success probability when trying to distinguish two states that are chosen uniformly at random. The latter has an operational meaning as the maximum probability that a purification of one state could pass a test for being a purification of the other (this is known as Uhlmann’s transition probability [Uhl76]). These distinguishability measures have generalizations to quantum channels, in the form of the diamond distance [Wat09b, VW16] for reviews of quantum computational hardness results of [Wat02b, RW05, Wat09c] lend credence to the belief that estimating these quantities efficiently is an interesting open question that we leave for future research. Already in [WZC21], it was shown that estimating the fidelity of two quantum states is possible in quantum polynomial time when one of the states is low rank, and the same is the case for estimating the trace distance under certain promises [WGL22, WZ23]. See also [CPCC20, CSZW22, TV21] for variational algorithms that estimate fidelity of states and [CSZW22, LLSL21] for variational algorithms to estimate trace distance. It is open to determine precise conditions under which estimation is possible for channel and strategy distinguishability measures.

We perform noiseless and noisy simulations of several of the algorithms provided. We find that in the noiseless scenario, all algorithms converge, for the examples considered, to the true known value of the distinguishability measure under consideration. In the noisy simulations, the algorithms converge well, and the parameters obtained exhibit a noise resilience, as put forward in [SKCC20]; i.e., the relevant quantity can be accurately estimated by inputting the parameters learned from the noisy simulator into the noiseless simulator.

Lastly, we discuss the computational complexity of various distance estimation algorithms. We prove that several fidelity and distance estimation algorithms are complete for well-known quantum complexity classes (see [Wat09a, VW16] for reviews of quantum computational complexity theory). In particular, we prove that estimating the fidelity between two pure states, a mixed state and a pure state, and estimating the Hilbert–Schmidt distance of two mixed states are BQP-complete problems. These aforementioned results follow by demonstrating that there is an efficient quantum algorithm for these tasks and by showing a reduction from an arbitrary BQP algorithm to one for these tasks. Thus, if we believe that there is a separation between the computational power of classical and quantum computers, then these estimation problems are those for which a quantum computer has an advantage. Several BQP-complete promise problems are known, including approximating the Jones polynomial.
In the rest of the paper, we provide details of the algorithms and results mentioned above. In particular, our paper proceeds as follows:

1. The various subsections of Section II are about estimating the fidelity of states, channels, and strategies. We begin in Section II A by establishing two quantum algorithms for estimating the fidelity of pure states, one of which is based on a state overlap test (Algorithm 1) and another that employs Bell state preparation and measurement along with a controlled unitary (Algorithm 2).

2. In Section II B, we generalize Algorithm 1 to estimate the fidelity of a pure state and a mixed state (see Algorithm 3).

3. In Section II C, we establish several quantum algorithms for estimating the fidelity of arbitrary states. Algorithm 4 generalizes Algorithm 2. Algorithm 5 generalizes the well-known swap test to the case of arbitrary states. Algorithm 6 is a variational algorithm that employs Bell measurements, as a generalization of the approach in [GECP13, SCC19] for pure states. Algorithm 7 is another variational algorithm that attempts to simulate a fidelity-achieving measurement, such as the Fuchs–Caves measurement [FC95], in order to estimate the fidelity.

4. In Section II D, we generalize Algorithm 4 to a quantum algorithm for estimating the fidelity of quantum channels (see Algorithm 8). This algorithm involves interaction with competing quantum provers, and interestingly, its acceptance probability is directly related to the fidelity of channels, thus giving the latter an operational meaning. Later, we replace the provers with parameterized circuits and arrive at a method for estimating the fidelity of channels.

5. In Section II E, we generalize the aforementioned approach in order to estimate the fidelity of strategies (a strategy is a sequence of quantum channels and thus generalizes the notion of a quantum channel).

6. In Section II F, we briefly discuss alternative methods for estimating the fidelity of channels and strategies, based on the approaches from Section II C for estimating the fidelity of states.

7. Section II G introduces a method for estimating the maximum output fidelity of two quantum channels, which has an application to generating a fixed point of a quantum channel (as discussed later on in Section VI).

8. In Sections II H and III, we generalize the whole development above to the case of testing similarity of arbitrary ensembles of states, channels, or strategies. We find that the acceptance probability of the corresponding algorithms is related to the secrecy measure from [KRS09], which can be understood as a measure of similarity of the states in an ensemble. We then establish generalizations of this measure for an ensemble of channels and an ensemble of strategies and remark how this has applications in private quantum reading [BDW18, DBW20].

9. We then move on in Section III to estimating trace-distance-based measures, for states, channels, and strategies. We stress that these various algorithms were already known, and our goal here is to investigate their performance using a variational approach. In Sections III A, III B, and III C, Algorithms 14, 15, and 16 provide methods for estimating the trace distance of states, the diamond distance of channels, and the strategy distance of strategies, respectively.

10. In Section III D, we provide two different but related algorithms for estimating the minimum trace distance between two quantum channels. The related approaches employ competing provers to do so.

11. In Section III E, we generalize the whole development for trace-distance based algorithms to the case of multiple states, channels, and strategies.

12. In Section IV, we discuss the results of numerical simulations of Algorithms 4–8, Algorithms 14–15, and Algorithm 19. We use both noiseless and noisy quantum simulators and a variational approach with parameterized circuits.

13. In Section V, we prove that the problems of evaluating the fidelity between two pure states, a pure state and a mixed state, and evaluating the Hilbert–Schmidt distance of two mixed states are BQP-complete (Theorem 12, 13, 14). We then
show that the problem of evaluating the fidelity between a channel with arbitrary input and a pure state is QMA-complete (Theorem 16). Finally, we demonstrate that the problem of evaluating the fidelity between a channel with separable input and a pure state is QMA(2)-complete (Theorem 17).

14. In Section VI, we discuss how Algorithm 10 can generate a fixed-point state or an approximate fixed-point state of a quantum channel.

We finally conclude in Section VII with a summary and some open questions.

II. ESTIMATING FIDELITY

In this section, we propose algorithms for several different fidelity problems. A summary of all algorithms presented in this section is available in Table I.

A. Estimating fidelity of pure states

We begin by outlining two simple quantum algorithms for estimating fidelity when both states are pure. A standard approach for doing so is to use the swap test [BBD+97, BCWdW01] or Bell measurements [GECP13, SCC19]. The approaches that we discuss below are different from these approaches. The first algorithm is a special case of that proposed in [Wat02b] (see also [CSZW22]), as well as a special case of Algorithm 3 presented later. The second algorithm involves a Bell-state preparation and projection, as well as controlled interactions, and it is a special case of Algorithm 4 presented later. We list both of these algorithms here for completeness and because later algorithms build upon them.

Suppose that the goal is to estimate the fidelity of pure states \( \psi^0 \) and \( \psi^1 \), and we are given access to quantum circuits \( U^0 \) and \( U^1 \) that prepare these states when acting on the all-zeros state. We now detail a first quantum algorithm for estimating the fidelity

\[
F(\psi^0, \psi^1) := |\langle \psi^1 | \psi^0 \rangle|^2. \tag{1}
\]

**Algorithm 1** The algorithm proceeds as follows:

1. Act with the circuit \( U^0 \) on the all-zeros state \( |0\rangle \).
2. Act with \( U^{1\dagger} \) and perform a measurement of all qubits in the computational basis.
3. Accept if and only if the all-zeros outcome is observed.

Algorithm 1 is depicted in Figure 1. The acceptance probability of Algorithm 1 is precisely equal to \(|\langle 0 | U^{1\dagger} U^0 | 0 \rangle|^2\), which by definition is equal to the fidelity in (1). In fact, Algorithm 1 is a quantum computational implementation of the well known operational interpretation of the fidelity as the probability that the state \( \psi^0 \) passes a test for being the state \( \psi^1 \).

Our next quantum algorithm for estimating fidelity makes use of a Bell-state preparation and projection. Its acceptance probability is equal to

\[
\frac{1}{2} \left( 1 + \sqrt{F(\psi^0, \psi^1)} \right) \tag{2}
\]

and thus gives a way to estimate the fidelity through repetition. It is a variational algorithm that optimizes over a phase \( \phi \) and makes use of the fact that

\[
\max_{\phi \in [0, 2\pi]} \Re[e^{i\phi} \langle \psi^0 | \psi^1 \rangle] = |\langle \psi^0 | \psi^1 \rangle|. \tag{3}
\]

This can be seen from the fact that the optimal phase \( \phi \) picked is such that

\[
e^{i\phi} = \frac{\langle \psi^1 | \psi^0 \rangle}{|\langle \psi^1 | \psi^0 \rangle|}. \tag{4}
\]

Let \( S \) denote the quantum system in which the states \( \psi^0 \) and \( \psi^1 \) are prepared.

**Algorithm 2** The algorithm proceeds as follows:

1. Prepare a Bell state

\[
|\Phi\rangle_{T-T} := \frac{1}{\sqrt{2}}(|00\rangle_T + |11\rangle_T) \tag{5}
\]

on registers \( T' \) and \( T \) and prepare system \( S \) in the all-zeros state \( |0\rangle_S \).

2. Using the circuits \( U^0_S \) and \( U^1_S \), perform the following controlled unitary:

\[
\sum_{i \in \{0, 1\}} |i\rangle_i T \otimes U^i_S, \tag{6}
\]

3. Act with the following unitary on system \( T' \):

\[
\begin{bmatrix}
1 & 0 \\
0 & e^{i\phi}
\end{bmatrix}. \tag{7}
\]
| Problem | Algorithms | Approach | Comparison |
|---------|------------|----------|------------|
| \(F(\psi, \phi)\) | Algorithm 1 | State Overlap | Algorithm 1 is simpler than Algorithm 2. Algorithm 2 generalizes in a straightforward manner to testing fidelity of mixed states. |
| \(F(\psi, \rho)\) | Algorithm 2 | Bell-State Overlap | - |
| \(F(\rho_0, \rho_1)\) | Algorithm 3 | State Overlap | - |
| \(F(N_0, N_1)\) | Algorithm 4 | Bell-State Overlap | Algorithm 4 is a generalization of Algorithm 2 for mixed state inputs. Algorithm 5 uses a controlled SWAP gate to generalize the SWAP Test. Requires more qubits, but no controlled unitaries to generate the states being tested. Algorithm 6 uses a variational unitary on the reference system of one state only. Algorithms 4, 5 and 6 are based on learning the Uhlmann unitary and provides a lower bound. Algorithm 7 is based on learning the optimal Fuchs–Caves measurement and provides an upper bound. |
| \(F(N_0, N_1)\) | Algorithm 5 | Generalized SWAP Test | - |
| \(F(N_0, N_1)\) | Algorithm 6 | Bell Measurement | - |
| \(p_{\text{aim}}(\{p(x), \rho^x\}_{x \in X})\) | Algorithm 7 | Fuchs–Caves Measurement | - |
| \(p_{\text{aim}}(\{p(x), N^x\}_{x \in X})\) | Algorithm 8 | Bell-State Overlap | Generalization of Algorithm 4 to ensemble of states. |
| \(p_{\text{aim}}(\{p(x), N^x\}_{x \in X})\) | Algorithm 9 | Bell-State Overlap | Generalization of Algorithm 8 to ensemble of channels. |
| \(p_{\text{aim},\text{max}}(\{p(x), N^x\}_{x \in X})\) | Algorithm 10 | Bell-State Overlap | Generalization of Algorithm 10 to ensemble of channels. |
| \(p_{\text{aim},\text{max}}(\{p(x), N^x\}_{x \in X})\) | Algorithm 11 | Bell-State Overlap | Generalization of Algorithm 10 to ensemble of channels. |

**TABLE I.** List of fidelity problems and algorithms addressed in this work. Approach used for each algorithm and comparison within a type of fidelity problem is also presented.

4. Perform a Bell measurement

\[
\{\Phi_{T'T}, I_{T'T} - \Phi_{T'T}\}
\]

on systems \(T'\) and \(T\). Accept if and only if the outcome \(\Phi_{T'T}\) occurs.

Figure 2 depicts Algorithm 2. After Step 3 of Algorithm 2, the overall state is as follows:

\[
\frac{1}{\sqrt{2}} \sum_{j \in \{0, 1\}} |jj\rangle_{T'T} e^{i j \phi} |\psi^j\rangle_S,
\]

and the acceptance probability is equal to

\[
\left\| \left( \frac{1}{\sqrt{2}} \sum_{j \in \{0, 1\}} |jj\rangle_{T'T} e^{i j \phi} |\psi^j\rangle_S \right) \right\|_2^2 = \frac{1}{4} \left\| \sum_{j \in \{0, 1\}} e^{i j \phi} |\psi^j\rangle_S \right\|_2^2
\]

\[
= \left\| \sum_{j \in \{0, 1\}} e^{i j \phi} |\psi^j\rangle_S \right\|_2^2 = \frac{1}{4} \left( 2 + 2 \text{Re}[e^{i \phi} \langle \psi^0 | \psi^1 \rangle] \right).
\]

**FIG. 2.** This figure depicts Algorithm 2 for estimating the fidelity of pure states generated by quantum circuits \(U^0\) and \(U^1\). The third gate with \(U^i\) in the box is defined in (6).
By choosing the optimal phase $\phi$ in (3), we find that the acceptance probability is equal to the expression in (2). Note that, through repetition, we can execute Algorithm 2 in a variational way to learn the optimal value of $\phi$.

Later on, in Section V, we prove that a promise version of the problem of estimating the fidelity of a pure state and a mixed state is a BQP-complete promise problem.

### B. Estimating fidelity when one state is pure and the other is mixed

In this section, we outline a simple quantum algorithm that estimates the fidelity between a mixed state $\rho_S$ and a pure state $\psi_S$. It is a straightforward generalization of Algorithm 1.

Let $U^\rho_{RS}$ be a quantum circuit that generates a purification $\varphi_{RS}$ of $\rho_S$ when acting on the all-zeros state of systems $RS$, and let $U^\psi_S$ be a circuit that generates $\psi_S$ when acting on the all-zeros state.

**Algorithm 3** The algorithm proceeds as follows:

1. Act on the all-zeros state $|0\rangle_{RS}$ with the circuit $U^\rho_{RS}$.
2. Act with $U^\psi_S$ on system $S$ and perform a measurement of all qubits of system $S$ in the computational basis.
3. Accept if and only if the all-zeros outcome is observed.

Figure 3 depicts Algorithm 3. The acceptance probability of Algorithm 3 is equal to the fidelity $F(\psi, \rho) = \langle \psi | \rho | \psi \rangle$, which follows because

$$
\| |0\rangle_S U^\psi_S U^\rho_{RS} |0\rangle_R \|_2^2 = \text{Tr}[(I_R \otimes |\psi\rangle \langle \psi|_S) |\varphi\rangle \langle \varphi|_{RS}] = \text{Tr}[|\psi\rangle \langle \varphi|_S \rho_S] = \langle \psi | \rho | \psi \rangle.
$$

We note here that it is not strictly necessary to have access to the reference system $R$ of $|\varphi\rangle_{RS}$ in order to execute Algorithm 3. It is only necessary to have some method of generating the reduced state $\rho_S$.

Later on, in Section V, we prove that a promise version of the problem of estimating the fidelity of a pure state and a mixed state is a BQP-complete promise problem.

### C. Estimating fidelity of arbitrary states

In this section, we outline several quantum algorithms for estimating the fidelity of arbitrary states on a quantum computer, some of which involve an interaction with a quantum prover (more precisely, the algorithms involving interaction with a prover are QSZK algorithms, where QSZK stands for “quantum statistical zero knowledge” [Wat02b, Wat09c]). The algorithms are different from the algorithm proposed in [Wat02b] (as also considered in [CSZW22]), which is based on Uhlmann’s formula for fidelity [Uhl76].

Suppose that the goal is to estimate the fidelity of states $\rho^0_S$ and $\rho^1_S$, defined as [Uhl76]

$$
F(\rho^0_S, \rho^1_S) := \left\| \sqrt{\rho^0_S} \sqrt{\rho^1_S} \right\|_1^2,
$$

where the trace norm of an operator $A$ is defined as $||A||_1 := \text{Tr}[\sqrt{A^+ A}]$. Suppose also that we are given access to quantum circuits $U^0_{RS}$ and $U^1_{RS}$ that prepare purifications $\psi^0_{RS}$ and $\psi^1_{RS}$ of $\rho^0_S$ and $\rho^1_S$, respectively, when acting on the all-zeros state $|0\rangle_{RS}$. Let us recall Uhlmann’s formula for fidelity [Uhl76]:

$$
F(\rho^0_S, \rho^1_S) = \max_{|\psi^0\rangle_{RS}, |\psi^1\rangle_{RS}} |\langle \psi^1 | \psi^0 \rangle_{RS}|^2,
$$

where the optimization is over all purifications $\psi^0_{RS}$ and $\psi^1_{RS}$ of $\rho^0_S$ and $\rho^1_S$, respectively. We note here that the fidelity can be computed by means of a semi-definite program [Wat13]. Also, the promise version of this problem, involving descriptions of quantum circuits as input, is a QSZK-complete promise problem [Wat02b], where QSZK stands for quantum statistical zero knowledge (see [Wat02b, Wat09c] for details of this complexity class). Thus, it is unlikely that anyone will find a general-purpose efficient quantum algorithm for estimating fidelity (i.e., one that does not involve interaction with an all-powerful prover).

We note that the algorithms in this subsection need the purification of the state of interest to be provided. In scenarios where the purification of a state is not available, there exist variational algorithms to learn the purification [EBS+23, CSZW22].
1. Controlled unitary and Bell state overlap

We now detail a QSZK algorithm for estimating the following quantity:

\[
\frac{1}{2} \left( 1 + \sqrt{F(\rho_S^0, \rho_S^1)} \right) . \tag{18}
\]

It is a QSZK algorithm because, in the case that the fidelity \(\sqrt{F(\rho_S^0, \rho_S^1)} \approx 1\), the verifier does not learn anything by interacting with the prover (i.e., the verifier only learns that the algorithm accepts with high probability). This algorithm is somewhat similar to the quantum algorithm proposed in [CHM\textsuperscript{+}16], which was used for estimating a quantity known as fidelity of recovery \([SW15]\). It is also similar to the algorithm described in Figure 3 of [KW00]. It can be understood as a generalization of Algorithm 2 from pure states to arbitrary states.

**Algorithm 4** The algorithm proceeds as follows:

1. The verifier prepares a Bell state

\[
|\Phi\rangle_{T'T'} := \frac{1}{\sqrt{2}}(|00\rangle_{T'T} + |11\rangle_{T'T}) \tag{19}
\]

on registers \(T'\) and \(T\) and prepares systems \(RS\) in the all-zeros state \(|0\rangle_{RS}\).

2. Using the circuits \(U_{RS}^0\) and \(U_{RS}^1\), the verifier performs the following controlled unitary:

\[
\sum_{i \in \{0,1\}} |i\rangle|i\rangle_T \otimes U_{RS}^i . \tag{20}
\]

3. The verifier transmits systems \(T'\) and \(R\) to the prover.

4. The prover prepares a system \(F\) in the \(|0\rangle_F\) state and acts on systems \(T', R,\) and \(F\) with a unitary \(P_{T'RF_{\rightarrow T'F'F'}}\) to produce the output systems \(T''\) and \(F''\), where \(T''\) is a qubit system.

5. The prover sends system \(T''\) to the verifier, who then performs a Bell measurement

\[
\{\Phi_{T''T}, I_{T''T} - \Phi_{T''T}\} \tag{21}
\]

on systems \(T''\) and \(T\). The verifier accepts if and only if the outcome \(\Phi_{T''T}\) occurs.

Figure 4 depicts Algorithm 4.

**Theorem 1** The acceptance probability of Algorithm 4 is equal to

\[
\frac{1}{2} \left( 1 + \sqrt{F(\rho_S^0, \rho_S^1)} \right) . \tag{22}
\]

**Proof.** The proof can be found in Appendix A 1.

2. Generalized swap test

We now detail another quantum algorithm for estimating the fidelity of arbitrary states, which is a generalization of the well-known swap test from [BBD+97, BCWdW01]. We note that this algorithm was used in [KW00, Figure 3] as part of their proof that QIP = QIP(3). A key difference between Algorithm 5 and [KW00, Figure 3] is that Algorithm 5 accepts if and only if both qubits at the end are measured to be in the all-zeros state, whereas it is written in [KW00, Figure 3] that their algorithm accepts if and only if the first qubit is measured to be in the zero state.

**Algorithm 5** The algorithm proceeds as follows:

1. The verifier prepares a Bell state

\[
|\Phi\rangle_{T'T'} := \frac{1}{\sqrt{2}}(|00\rangle_{T'T} + |11\rangle_{T'T}) \tag{23}
\]

on registers \(T'\) and \(T\) and prepares systems \(R_iS_iR_2S_2\) in the all-zeros state \(|0\rangle_{R_iS_iR_2S_2}\).

2. Using the circuits \(U_{RS}^0\) and \(U_{RS}^1\), the verifer acts on \(R_iS_iR_2S_2\) to prepare the two pure states \(|\psi^{S_i}⟩_{R_iS_i}\) and \(|\psi^{S_2}⟩_{R_2S_2}\).

3. The verifier performs a controlled SWAP from qubit \(T\) to systems \(S_1\) and \(S_2\), which applies the identity if the control qubit is \(|0\rangle\) and swaps \(S_1\) with \(S_2\) if the control qubit is \(|1\rangle\).

4. The verifier transmits systems \(T'\), \(R_1\), and \(R_2\) to the prover.

5. The prover prepares a system \(F\) in the \(|0\rangle_F\) state and acts on systems \(T'', R_1, R_2,\) and \(F\) with a unitary \(P_{T'R_1R_2F_{\rightarrow T''F''F'}}\) to produce the output systems \(T''\) and \(F''\), where \(T''\) is a qubit system.

6. The prover sends system \(T''\) to the verifier, who then performs a Bell measurement

\[
\{\Phi_{T''T}, I_{T''T} - \Phi_{T''T}\} \tag{24}
\]

on systems \(T''\) and \(T\). The verifier accepts if and only if the outcome \(\Phi_{T''T}\) occurs.

Figure 5 depicts Algorithm 5.

**Theorem 2** The acceptance probability of Algorithm 5 is equal to

\[
\frac{1}{2} \left( 1 + F(\rho_S^0, \rho_S^1) \right) . \tag{25}
\]

**Proof.** The proof can be found in Appendix A 2.
FIG. 4. This figure depicts Algorithm 4 for estimating the fidelity of mixed states generated by quantum circuits $U^0_{RS}$ and $U^1_{RS}$.

FIG. 5. This figure depicts Algorithm 5 for estimating the fidelity of mixed states generated by quantum circuits $U^0_{RS}$ and $U^1_{RS}$. Algorithm 5 represents a generalization of the well known swap test for estimating the fidelity of pure states.

3. Variational algorithm with Bell measurements

A third method for estimating the fidelity of arbitrary multi-qubit states is a variational algorithm that is based on a generalization of the approach outlined in [GECP13, SCC19]. The approach from [GECP13, SCC19] employs Bell measurements to estimate the expectation of the SWAP observable, which in turn allows for estimating the fidelity of multi-qubit pure states. See also [Bru04].

We begin in this section by recalling the basic idea from [GECP13, SCC19] for estimating fidelity of pure states. Let $\psi_S$ and $\varphi_S$ be $m$-qubit pure states of a system $S$ (so that $S = S_1 \cdots S_m$, where each $S_i$ is a qubit system, for $i \in \{1, \ldots, m\}$). Let $F_{S\tilde{S}}$ denote the unitary operator that swaps systems $S$ and $\tilde{S}$, and recall that

$$\operatorname{Tr}[F_{S\tilde{S}}(\psi_S \otimes \varphi_{\tilde{S}})] = |\langle \psi | \varphi \rangle|^2 = F(\psi_S, \varphi_S).$$ (26)

Consider that

$$F_{S\tilde{S}} = F_{S_1\tilde{S}_1} \otimes F_{S_2\tilde{S}_2} \otimes \cdots \otimes F_{S_m\tilde{S}_m}.$$ (27)

Now observe that

$$F_{S_i\tilde{S}_i} = \sum_{x,z \in \{0,1\}} (-1)^{xz} \Phi_{x,z}^{S_i\tilde{S}_i},$$ (28)

where the Bell states are defined as

$$|\Phi^{0,0}_{0,0}\rangle := \frac{1}{\sqrt{2}} (|00\rangle + |11\rangle),$$ (29)
$$|\Phi^{0,1}_{0,1}\rangle := \frac{1}{\sqrt{2}} (|00\rangle - |11\rangle),$$ (30)
$$|\Phi^{1,0}_{1,0}\rangle := \frac{1}{\sqrt{2}} (|01\rangle + |10\rangle),$$ (31)
$$|\Phi^{1,1}_{1,1}\rangle := \frac{1}{\sqrt{2}} (|01\rangle - |10\rangle).$$ (32)

We then conclude that

$$F(\psi_S, \varphi_S) = \operatorname{Tr} \left[ \bigotimes_{i=1}^m F_{S_i\tilde{S}_i} \right] (\psi_S \otimes \varphi_S)$$ (33)

$$= \operatorname{Tr} \left[ \bigotimes_{i=1}^m \sum_{x,z \in \{0,1\}} (-1)^{xz} \Phi_{x,z}^{S_i\tilde{S}_i} \right] (\psi_S \otimes \varphi_S)$$ (34)
Thus, the approach of [GECP13, SCC19] is to estimate \( F(\psi_S, \psi_S) \) by repeatedly performing Bell measurements on corresponding qubits of \( \psi_S \) and \( \varphi_S \) followed by classical postprocessing of the outcomes. In particular, for \( j \in \{1, \ldots, n\} \), set \( Y_j = (-1)^{\sum_{i=1}^{m} x_i z_i} \), where \( x_1, z_1, \ldots, x_m, z_m \in \{0,1\} \) are the outcomes of the Bell measurements on the \( j \)th iteration. Then set \( \overline{Y}^n := \frac{1}{n} \sum_{j=1}^n Y_j \). By the Hoeffding inequality [Hoe63], for accuracy \( \varepsilon \in (0,1) \) and failure probability \( \delta \in (0,1) \), we are guaranteed that
\[
\text{Pr}[|\overline{Y}^n - F(\psi_S, \psi_S)| \leq \varepsilon] \geq 1 - \delta,
\]
as long as \( n \geq \frac{2}{\varepsilon^2} \ln \left( \frac{2}{\delta} \right) \). Thus, the algorithm is polynomial in the inverse accuracy and logarithmic in the inverse failure probability.

We now form a simple generalization of this algorithm to estimate the fidelity of arbitrary states \( \rho_S^0 \) and \( \rho_S^1 \), in which we perform a variational optimization over unitaries that act on the reference system of one of the states. For \( i \in \{0,1\} \), let \( U_{RS} \) be an \( m \)-qubit unitary that acts on \( |0\rangle_{RS} \) to generate the \( m \)-qubit state \( |\psi^{i}\rangle_{RS} \); i.e.,
\[
|\psi^{i}\rangle_{RS} = U_{RS} |0\rangle_{RS},
\]
such that
\[
\rho_S^{i} = \text{Tr}_R[|\psi^{i}\rangle_\psi^{i}\rangle_{RS}].
\]

**Algorithm 6** Set the error tolerance \( \varepsilon > 0 \). Set \( \eta, \delta \in (0,1) \). The algorithm proceeds as follows:

1. Prepare systems \( R_1 S_1 R_2 S_2 \) in the all-zeros state \( |0\rangle_{R_1 S_1 R_2 S_2} \).
2. Act with the circuits \( U_{RS}^0 \) and \( U_{RS}^1 \) on systems \( R_1 S_1 R_2 S_2 \) to prepare the two pure states \( |\psi^{0}\rangle_{R_1 S_1} \) and \( |\psi^{1}\rangle_{R_2 S_2} \).
3. Perform a unitary \( V_R(\theta) \) on system \( R_1 \).
4. For \( j \in \{1, \ldots, n\} \), where \( n \geq \frac{2}{\varepsilon^2} \ln \left( \frac{2}{\delta} \right) \), for \( i \in \{0,1\} \), perform a Bell measurement on qubit \( i \) of system \( R_1 \) and qubit \( i \) of system \( R_2 \), with outcomes \( x_h^i \) and \( z_h^i \), and perform a Bell measurement on qubit \( i \) of system \( S_1 \) and qubit \( i \) of system \( S_2 \), with outcomes \( x_s^i \) and \( z_s^i \). Set \( Y_{j}(\theta) = (-1)^{\sum_{i=1}^{m} x_h^i z_h^i + x_s^i z_s^i} \).

5. Set
\[
\overline{Y}^n(\theta) := \frac{1}{n} \sum_{j=1}^n Y_{j}(\theta),
\]
as an estimate of
\[
F_{\theta} = \left| \langle \psi^{i}\rangle_{RS} V_R(\theta) \otimes I_S |\psi^{i}\rangle_{RS} \right|^2,
\]
so that
\[
\text{Pr}[|\overline{Y}^n(\theta) - F_{\theta}| \leq \eta] \geq 1 - \delta.
\]

6. Perform a maximization of the reward function \( \overline{Y}^n(\theta) \) and update the parameters in \( \theta \).

7. Repeat 1-6 until the reward function \( \overline{Y}^n(\theta) \) converges with tolerance \( \varepsilon \), so that \( |\Delta \overline{Y}^n(\theta)| \leq \varepsilon \), or until some maximum number of iterations is reached. (Here \( \Delta \overline{Y}^n(\theta) \) represents the difference in \( \overline{Y}^n(\theta) \) from the previous and current iteration.)

8. Output the final \( \overline{Y}^n(\theta) \) as an estimate of the fidelity \( F(\rho_S^0, \rho_S^1) \).

Figure 6 depicts Algorithm 6. Since this is a variational algorithm, it is not guaranteed to converge or have a specified runtime, other than running for a maximum number of iterations. However, it is clearly a generalization of the algorithm from [GECP13, SCC19], in which we estimate the fidelity
\[
\left| \langle \psi^{i}\rangle_{RS} V_R(\theta) \otimes I_S |\psi^{i}\rangle_{RS} \right|^2 = F(\rho_S^{i}, V_R(\theta) \rho_S^{i} V_R(\theta)^{\dagger})
\]
at each iteration of the algorithm. If we could actually optimize over all possible unitaries acting on the reference system \( R \), then the algorithm would indeed estimate the fidelity, as a consequence of Uhlmann’s theorem [Uhl76]:
\[
F(\rho_S^{0}, \rho_S^{1}) = \sup_{V_R} F(\rho_S^{0}, V_R \rho_S^{0} V_R^{\dagger}).
\]
However, by optimizing over only a subset of all unitaries, Algorithm 6 estimates a lower bound on the fidelity $F(\rho_S^0, \rho_S^1)$.

4. Variational algorithm for Fuchs–Caves measurement

Algorithm 4 from Section II C 1 is based on Uhlmann’s formula for fidelity in (17), and the same is true for Algorithm 5 from Section II C 2 and Algorithm 6 from Section II C 3. An alternate optimization formula for the fidelity of states $\rho_S^0$ and $\rho_S^1$ is as follows [FC95]:

$$F(\rho_S^0, \rho_S^1) = \left[ \min_{\{ \Lambda_x^s \}_x} \sum_x \sqrt{\text{Tr}[\Lambda_x^s \rho_S^0 \text{Tr}[\Lambda_x^s \rho_S^1]]} \right]^2,$$  \hspace{1cm} (45)

where the minimization is over every positive operator-valued measure $\{ \Lambda_x^s \}_x$ (i.e., the operators satisfy $\Lambda_x^s \geq 0$ for all $x$ and $\sum_x \Lambda_x^s = I_S$). A measurement achieving the optimal value of the fidelity is known as the Fuchs–Caves measurement [FC95] and has the form $\{|\varphi_x\rangle \langle \varphi_x|\}_x$, where $|\varphi_x\rangle$ is an eigenvector, with eigenvalue $\lambda_x$, of the following operator geometric mean of $\rho^0$ and $\rho^1$ (also called “quantum likelihood ratio” operator in [Fuc96]):

$$M := (\rho^1)^{-1/2} \sqrt{(\rho^1)^{1/2} \rho^0 (\rho^1)^{1/2} (\rho^1)^{-1/2}},$$  \hspace{1cm} (46)

so that

$$M = \sum_x \lambda_x |\varphi_x\rangle \langle \varphi_x|.$$  \hspace{1cm} (47)

That is, it is known from [FC95, Fuc96] that

$$F(\rho_S^0, \rho_S^1) = \left[ \sum_x \sqrt{\text{Tr}[|\varphi_x\rangle \langle \varphi_x| \rho_S^0 \text{Tr}[|\varphi_x\rangle \langle \varphi_x| \rho_S^1]]} \right]^2.$$  \hspace{1cm} (48)

Thus, we can build a variational algorithm around this formulation of fidelity, with the idea being to optimize over parameterized measurements in an attempt to optimize the fidelity, while at the same time learn the Fuchs–Caves measurement (or a different fidelity-achieving measurement). In contrast to the other variational algorithms presented in previous sections, this alternate approach leads to an upper bound on the fidelity.

Before detailing the algorithm, recall the Naimark extension theorem [Nai40] (see also [Wil17, Wat18, KW20]), which states that a general POVM $\{\Lambda_x^s\}_x$ with $m$ outcomes, acting on a quantum state $\rho$ of a $d$-dimensional system $S$, can be realized as a unitary interaction $U_{SP}$ of the system $S$ with an $m$-dimensional probe system $P$, followed by a projective measurement $\{|x\rangle \langle x|\}_x$ acting on the probe system. That is,

$$\text{Tr}[\Lambda_x^s \rho_S] = \text{Tr}[\rho_S \otimes |x\rangle \langle x|_P] U_{SP}(\rho_S \otimes |0\rangle \langle 0|_P) U_{SP}^\dagger.$$  \hspace{1cm} (49)

It suffices to choose $U_{SP}$ so that

$$U_{SP}|\psi\rangle_S |0\rangle_P = \sum_x \sqrt{\lambda_x} |\psi\rangle_S |x\rangle_P.$$  \hspace{1cm} (50)

Thus, we can express the optimization problem in (45) as follows:

$$\sqrt{F(\rho_S^0, \rho_S^1)} = \min_{U_{SP}} \left[ \sum_x \text{Tr}[\rho_S \otimes |x\rangle \langle x|_P U_{SP}(\rho_S^0 \otimes |0\rangle \langle 0|_P) U_{SP}^\dagger \times \text{Tr}[\rho_S \otimes |x\rangle \langle x|_P U_{SP}(\rho_S^1 \otimes |0\rangle \langle 0|_P) U_{SP}^\dagger]\right].$$  \hspace{1cm} (51)

By replacing the optimization in (51) over all unitaries with an optimization over parameterized ones, we arrive at a variational algorithm for estimating fidelity:

**Algorithm 7** Set $n \in \mathbb{N}$ and the error tolerance $\varepsilon > 0$. The algorithm proceeds as follows:

1. For $j \in \{1, \ldots, n\}$, prepare system $S_1$ in the state $\rho_{S_1}^j$ and system $S_2$ in the state $\rho_{S_2}^j$, and prepare systems $P_1$ and $P_2$ in the all-zeros state $|0\rangle_{P_1} \otimes |0\rangle_{P_2}$.

2. Act with the circuit $U_{S_1, P_1} (\theta)$ on systems $S_1 P_1$ and act with the same circuit $U_{S_2, P_2} (\theta)$ on systems $S_2 P_2$.

3. Measure system $P_1$ in the computational basis and record the outcome as $y_j$, and measure system $P_2$ in the computational basis and record the outcome as $z_j$.

4. Using the measurement data $\{y_j\}_{j=1}^n$ and $\{z_j\}_{j=1}^n$, calculate the empirical distributions $\tilde{p}_\theta(x)$ and $\tilde{q}_\theta(x)$, where $\tilde{p}_\theta(x)$ is the empirical distribution resulting from

$$p_\theta(x) := \text{Tr}[U_S \otimes |x\rangle \langle x|_P U_{SP}(\theta) (\rho_S^0 \otimes |0\rangle \langle 0|_P) U_{SP}(\theta)]$$  \hspace{1cm} (52)

and $\tilde{q}_\theta(x)$ is the empirical distribution resulting from

$$q_\theta(x) := \text{Tr}[U_S \otimes |x\rangle \langle x|_P U_{SP}(\theta) (\rho_S^1 \otimes |0\rangle \langle 0|_P) U_{SP}(\theta)]$$  \hspace{1cm} (53)

5. Output

$$F(\tilde{p}_\theta, \tilde{q}_\theta) := \left[ \sum \sqrt{\tilde{p}_\theta(x) \tilde{q}_\theta(x)} \right]^2$$  \hspace{1cm} (54)

as an estimate of $F(p_\theta, q_\theta)$.  

6. Perform a minimization of the cost function $F(\tilde{p}_\theta, \tilde{q}_\theta)$ and update the parameters in $\theta$. 


7. Repeat 1-6 until the cost function \(F(\rho_0, \tilde{q}_0)\) converges with tolerance \(\varepsilon\), so that \(|\Delta F(\rho_0, \tilde{q}_0)| \leq \varepsilon\), or until some maximum number of iterations is reached. (Here \(\Delta F(\rho_0, \tilde{q}_0)\) represents the difference in \(F(\rho_0, \tilde{q}_0)\) from the previous and current iterations.)

8. Output the final value of \(F(\rho_0, \tilde{q}_0)\) as an estimate of the fidelity \(F(\rho_0^S, \rho_0^S)\).

Figure 7 depicts Algorithm 7. As before, since this is a variational algorithm, it is not guaranteed to converge or have a specified runtime, other than running for a maximum number of iterations. One advantage of this algorithm is that it does not require purifications of the states \(\rho_0^S\) and \(\rho_0^S\). All it requires is a circuit or method to prepare these states, and then it performs measurements on these states, in an attempt to learn an optimal measurement with respect to the cost function \(F(\rho_0, \tilde{q}_0)\).

In Algorithm 7, we did not specify how large \(n\) should be in order to get a desired accuracy of the estimator in (54) for the classical fidelity \(F(p_0, q_0)\). This estimator is called a “plug-in estimator” in the literature on this topic, and it is a biased estimator, which however converges to \(F(p_0, q_0)\) in the asymptotic limit \(n \to \infty\). As a consequence of the estimator in (54) being biased, the Hoeffding inequality does not readily apply in this case. As far as we can tell, it is an open question to determine the rate of convergence of this estimator to \(F(p_0, q_0)\). Related work on this topic has been considered in [GW05, Gut05, GW07, Gut09, GW13].

D. Estimating fidelity of channels

In this section, we outline a method for estimating the fidelity of channels on a quantum computer, by means of an interaction with competing quantum provers [GW05, Gut05, GW07, Gut09, GW13]. The goal of one prover is to maximize the acceptance probability, while the goal of the other prover is to minimize the acceptance probability. We refer to the first prover as the max-prover and the second as the min-prover. The specific setting that we deal with is called a double quantum interactive proof (DQIP) [GW13], due to the fact that the min-prover goes first and then the max-prover goes last. The class of promise problems that can be solved in this model is equivalent to PSPACE [GW13], which is the class of problems that can be decided on a classical computer with polynomial memory.

Let us recall that the fidelity of channels \(N^0_{A \to B}\) and \(N^1_{A \to B}\) is defined as follows [GLN05]:

\[
F(N^0_{A \to B}, N^1_{A \to B}) := \inf_{\rho_{RA}} F(N^0_{A \to B}(\rho_{RA}), N^1_{A \to B}(\rho_{RA})),
\]

where the infimum is over every state \(\rho_{RA}\), with the reference system \(R\) arbitrarily large. It is known that the infimum is achieved by a pure state \(\psi_{RA}\) with the reference system \(R\) isomorphic to the channel input system \(A\), so that

\[
F(N^0_{A \to B}, N^1_{A \to B}) := \min_{\psi_{RA}} F(N^0_{A \to B}(\psi_{RA}), N^1_{A \to B}(\psi_{RA})).
\]

It is also known that it is possible to calculate the fidelity of channels by means of a semi-definite program [YF17, KW21b], which provides a way to verify the output of our proposed algorithm for sufficiently small examples.

Suppose that the goal is to estimate the fidelity of channels \(N^0_{A \to B}\) and \(N^1_{A \to B}\), and we are given access to quantum circuits \(U^0_{AE' \to BE}\) and \(U^1_{AE' \to BE}\) that realize isometric extensions of the channels \(N^0_{A \to B}\) and \(N^1_{A \to B}\), respectively, in the sense that

\[
N^i_{A \to B}(\omega_A) = \text{Tr}_E[U^i_{AE' \to BE}(\omega_A \otimes |0\rangle\langle 0|_{E'}) (U^i_{AE' \to BE})^\dagger],
\]

for \(i \in \{0, 1\}\).

We now provide a DQIP algorithm for estimating the following quantity:

\[
\frac{1}{2} \left( 1 + \sqrt{F(N^0_{A \to B}, N^1_{A \to B})} \right),
\]

which is based in part on Algorithm 4 but instead features an optimization over input states of the min-prover.

Algorithm 8 The algorithm proceeds as follows:

1. The verifier prepares a Bell state

\[
|\Phi\rangle_T T' := \frac{1}{\sqrt{2}} (|00\rangle_{T' T} + |11\rangle_{T' T})
\]

on registers \(T'\) and \(T\) and prepares system \(E'\) in the all-zeros state \(|0\rangle_{E'}\).

2. The min-prover transmits the system \(A\) of the state \(|\psi\rangle_{RA}\) to the verifier.
3. Using the circuits $U_{AE→BE}^3$ and $U_{AE→BE}^1$, the verifier performs the following controlled unitary:

$$\sum_{i \in \{0,1\}} |i\rangle_i |T \rangle \otimes U_{AE→BE}^1.$$  \hspace{1cm} (60)

4. The verifier transmits systems $T'$ and $E$ to the max-prover.

5. The max-prover prepares a system $F$ in the $|0\rangle_F$ state and acts on systems $T'$, $E$, and $F$ with a unitary $P_{T'EF→T''F'}$ to produce the output systems $T''$ and $F'$, where $T''$ is a qubit system.

6. The max-prover sends system $T''$ to the verifier, who then performs a Bell measurement

$$\left\{\Phi_{T''T}, I_{T''T} - \Phi_{T''T}\right\}$$  \hspace{1cm} (61)

on systems $T''$ and $T$. The verifier accepts if and only if the outcome $\Phi_{T''T}$ occurs.

Figure 8 depicts Algorithm 8.

**Theorem 3** The acceptance probability of Algorithm 8 is equal to

$$\frac{1}{2} \left(1 + \sqrt{F(N_{A→B}^0, N_{A→B}^1)} \right).$$  \hspace{1cm} (62)

**Proof.** The proof can be found in Appendix A.3. ■

**Proposition 1** An alternative expression for the acceptance probability of Algorithm 8 is

$$\min_{\rho_{RA}} \max_{P_{T'E→T''}} \text{Tr}[\Phi_{T''T} P_{T'E→T''}(M_{A→T'TBE}(\rho_{RA}))] = \max_{P_{T'E→T''}} \min_{\rho_{RA}} \text{Tr}[\Phi_{T''T} P_{T'E→T''}(M_{A→T'TBE}(\rho_{RA}))],$$  \hspace{1cm} (63)

where $\rho_{RA}$ is a quantum state, $P_{T'E→T''}$ is a quantum channel, and $M_{A→T'TBE}$ is a quantum channel defined as

$$M_{A→T'TBE}(\rho_{RA}) := \frac{1}{2} \sum_{i,j \in \{0,1\}} |ii\rangle_i |jj\rangle_j |T \rangle \otimes (\rho_{RA} \otimes |0\rangle_0 |0\rangle_{E'}) (U^j)^\dagger.$$  \hspace{1cm} (64)

with $U^j \equiv U_{AE→BE}^1$.

**Proof.** In Step 2 of Algorithm 8, the min-prover could send a mixed quantum state $\rho_{RA}$ instead of sending a pure state. The acceptance probability does not change under this modification due to the argument around (55)–(56). Furthermore, due to the Stinespring dilation theorem [Sti55], the actions of tensoring in $|0\rangle_F$, performing the unitary $P_{T'EF→T''F'}$, and tracing over system $F'$ are equivalent to performing a quantum channel $P_{T'E→T''}$. Under these observations, consider that the acceptance probability is then equal to

$$\text{Tr}[\Phi_{T''T} P_{T'E→T''}(M_{A→T'TBE}(\rho_{RA}))],$$  \hspace{1cm} (65)

where the quantum channel $M_{A→T'TBE}$ is defined in (64). Performing the optimizations $\min_{\rho_{RA}} \max_{P_{T'E→T''}}$, then leads to the first expression in (63). Considering that the set of channels is convex and the set of states is convex, and the objective function in (65) is linear in $\rho_{RA}$ for fixed $P_{T'E→T''}$ and linear in $P_{T'E→T''}$ for fixed $\rho_{RA}$, the minimax theorem [Sio58] applies and we can exchange the optimizations. ■

Proposition 1 indicates that if the provers involved can optimize over all possible states and channels, then indeed the order of optimization can be exchanged. However, in a variational algorithm, the optimization is generally dependent upon the order in which it is conducted because we are not optimizing over all possible states and channels, but instead optimizing over parameterized circuits. In this latter case, the state space is no longer convex and the objective function no longer linear in these parameters. However, we can still attempt the following “see-saw” strategy in a variational algorithm: first minimize the objective function with respect to the input state $\psi_{RA}$ while keeping the unitary $P_{T'EF→T''F'}$ fixed. Then maximize the objective function with respect to the unitary $P_{T'EF→T''F'}$ while keeping the state $\psi_{RA}$ fixed. Then repeat this process some number of times. We consider this approach in Section IV.E.

**E. Estimating fidelity of strategies**

In this section, we extend Algorithm 8 beyond estimating the fidelity of channels to estimating the fidelity of general strategies [GRS18], by conducting several rounds of interaction with the min-prover followed by a single interaction with the max-prover at the end.

We now develop this idea in detail. Let us first recall the definition of a quantum strategy from [GW07, CDP08, CDP09, Gut09, Gut12, GRS18]. An $n$-turn quantum strategy $\mathcal{N}^{(n)}$, with $n \geq 1$, input systems $A_1, \ldots, A_n$, and output systems $B_1, \ldots, B_n$, consists of the following:

1. memory systems $M_1, \ldots, M_{n-1}$, and
2. quantum channels $\mathcal{N}_{A_1→M_1B_1}, \mathcal{N}_{M_2A_2→M_2B_2}, \ldots, \mathcal{N}_{M_{n-2}A_{n-1→M_{n-1}B_{n-1}}}, \text{and } \mathcal{N}_{M_{n-1}A_n→B_n}$.

It is implicit that any of the systems involved can be trivial systems, which means that state preparation and measurements are included as special cases. A co-strategy interacts with a strategy; co-strategies are in fact strategies also, but it is useful conceptually to provide an explicit means by which an agent can interact with a strategy. An $(n-1)$-turn co-strategy $\mathcal{S}^{(n-1)}$, with input systems $B_1, \ldots, B_n$ and output systems $A_1, \ldots, A_n$, consists of the following:
The result of the interaction of the strategy $N^{(n)}$ with the co-strategy $S^{(n-1)}$ is a quantum state on systems $R_n B_n$, and we employ the shorthand

$$N^{(n)} \circ S^{(n-1)}$$

(66)
to denote this quantum state. Figure 9 depicts a three-turn strategy interacting with a two-turn co-strategy.

Let $N^{0,(n)}$ and $N^{1,(n)}$ denote two compatible, $n$-turn quantum strategies, meaning that all systems involved in these strategies are the same but the channels that make up the strategies are possibly different. The fidelity of the strategies $N^{0,(n)}$ and $N^{1,(n)}$ is defined as [GRS18]

$$F(N^{0,(n)}, N^{1,(n)}) := \inf_{S^{(n-1)}} F(N^{0,(n)} \circ S^{(n-1)}, N^{1,(n)} \circ S^{(n-1)}),$$

(67)

where the optimization is over every co-strategy $S^{(n-1)}$. One can interpret the strategy fidelity in (67) as a generalization of the fidelity of channels in (55), in which the idea is to optimize the fidelity measure over all possible co-strategies that can be used to distinguish the strategies $N^{0,(n)}$ and $N^{1,(n)}$. It follows from a standard data-processing argument that it suffices to perform the optimization in (67) over co-strategies involving an initial pure state $\rho_{R_n A_1}$ and channels $S^1_{R_1 B_1 \rightarrow R_2 A_2}$, $S^2_{R_2 B_2 \rightarrow R_3 A_3}$, ..., and $S^{n-1}_{R_{n-1} B_{n-1} \rightarrow R_n A_n}$ that are each isometric channels (these are called pure co-strategies in [GRS18]). We also note here that the measure in (67) is generalized by the generalized strategy divergence of [WW19].

The goal of this section is to delineate a DQIP algorithm for estimating the fidelity of strategies $N^{0,(n)}$ and $N^{1,(n)}$. To do so, we suppose that the verifier has access to unitary circuits that realize isometric extensions of all channels involved in the strategies. That is, for $i \in \{0, 1\}$, there exists a unitary channel $U_{A_1 E'_i \rightarrow M_1 B_1 E_i}^{1}$ such that

$$\text{Tr}_{E_i} \left[ U_{A_1 E'_i \rightarrow M_1 B_1 E_i}^{1} (\rho_{A_1} \otimes |0\rangle \langle 0|_{E'_i}) \right]$$

(68)

for every input state $\rho_{A_1}$; for $j \in \{2, \ldots, n-1\}$, there exists a unitary channel $U_{M_{j-1} A_j E'_j \rightarrow M_j B_j E_j}^{j}$ such that...
The algorithm proceeds as follows:

1. The verifier prepares a Bell state

\[ |\Phi\rangle_{TT} := \frac{1}{\sqrt{2}} (|00\rangle_{TT} + |11\rangle_{TT}) \]  

for every input state \( \rho_{A_i} \); and there exists a unitary channel \( U_{M_{n-1}A_n}^{E_n} \rightarrow B_n E_n \) such that

\[ M_{n-1}^{A_n} A_n \rightarrow B_n (\rho_{A_n}) = \] 

\[ \text{Tr}_{E_n} \left( U_{M_{n-1}A_n}^{E_n} \rightarrow B_n E_n (\rho_{A_n} \otimes |00\rangle \langle 00|_{E_n}) \right), \]  

(69)

for every input state \( \rho_{A_n} \). We use the notation \( U_{M_{A_i}E_i}^{A_i} \rightarrow M_{B_i} E_i \), \( U_{M_{n-1}A_n}^{E_n} \rightarrow B_n E_n \), and \( U_{M_{n-1}A_n}^{E_n} \rightarrow B_n E_n \) to refer to the unitary circuits.

We now provide a DQIP algorithm for estimating the following quantity:

\[ \frac{1}{2} \left( 1 + \sqrt{F(N^{0,(n)}, N^{1,(n)})} \right), \]  

(71)

which is based in part on Algorithm 8 but instead features an optimization over all co-strategies of the min-prover.

**Algorithm 9**

The algorithm proceeds as follows:

1. The verifier prepares a Bell state

\[ |\Phi\rangle_{TT} := \frac{1}{\sqrt{2}} (|00\rangle_{TT} + |11\rangle_{TT}) \]  

(72)

on registers \( T' \) and \( T \) and prepares systems \( E_1' \cdots E_n' \) in the all-zeros state \( |00\rangle_{E_1' \cdots E_n'} \).

2. The min-prover transmits the system \( A \) of the state \( |\psi\rangle_{RA} \) to the verifier.

3. Using the circuits \( U_{A_1 E_1'} \rightarrow M_{B_1} E_1 \) and \( U_{M_{n-1}A_n}^{E_n} \rightarrow B_n E_n \), the verifier performs the following controlled unitary:

\[ \sum_{i \in \{0,1\}} |i\rangle\langle i| \otimes U_{A_1 E_1'} \rightarrow M_{B_1} E_1 \]  

(73)

4. The verifier transmits system \( B_1 \) to the min-prover, who subsequently acts with the isometric quantum channel \( S_{R_1 B_1} \rightarrow R_2 A_2 \) and then sends system \( A_2 \) to the verifier.

5. For \( j \in \{2, \ldots, n-1\} \), using the circuits \( U_{A_1 E_1'} \rightarrow M_{B_1} E_1 \), \( U_{M_{n-1}A_n}^{E_n} \rightarrow B_n E_n \), and \( U_{M_{n-1}A_n}^{E_n} \rightarrow B_n E_n \), the verifier performs the following controlled unitary:

\[ \sum_{i \in \{0,1\}} |i\rangle\langle i| \otimes U_{A_1 E_1'} \rightarrow M_{B_1} E_1 \]  

(74)

The verifier transmits system \( B_j \) to the min-prover, who subsequently acts with the isometric quantum channel \( S_{R_1 B_j} \rightarrow R_{j+1} A_{j+1} \) and then sends system \( A_{j+1} \) to the verifier.

6. Using the circuits \( U_{A_1 E_1}^{E_n} \rightarrow B_n E_n \) and \( U_{M_{n-1}A_n}^{E_n} \rightarrow B_n E_n \), the verifier performs the following controlled unitary:

\[ \sum_{i \in \{0,1\}} |i\rangle\langle i| \otimes U_{A_1 E_1}^{E_n} \rightarrow B_n E_n \]  

(75)

7. The verifier transmits systems \( T', E_1, \ldots, E_n \) to the max-prover.

8. The max-prover prepares a system \( F \) in the \( |00\rangle_F \) state and acts on systems \( T', E_1, \ldots, E_n, \) and \( F \) with a unitary \( P_T E_1 \cdots E_n F \rightarrow T'' F' \) to produce the output systems \( T'' \) and \( F' \), where \( T'' \) is a qubit system.

9. The max-prover sends system \( T'' \) to the verifier, who then performs a Bell measurement

\[ \{ \Phi_{T''T}, I_{T''T} - \Phi_{T''T} \} \]  

(76)

on systems \( T'' \) and \( T \). The verifier accepts if and only if the outcome \( \Phi_{T''T} \) occurs.

Figure 10 depicts Algorithm 9.

**Theorem 4**

The acceptance probability of Algorithm 9 is equal to

\[ \frac{1}{2} \left( 1 + \sqrt{F(N^{0,(n)}, N^{1,(n)})} \right), \]  

(77)

where \( \sqrt{F(N^{0,(n)}, N^{1,(n)})} \) is the strategy fidelity defined in (67).

**Proof.** The proof can be found in Appendix A 4. □

**F. Alternate methods of estimating the fidelity of channels and strategies**

We note briefly here that other methods for estimating fidelity of channels can be based on Algorithms 5, 6, and 7. It is not clear how to phrase them in the language of quantum interactive proofs, in such a way that the acceptance probability is a simple function of the channel fidelity. However, we can employ variational algorithms in which we repeat the circuit for determining an optimal input state \( \psi_{RA} \) for the channel fidelity. Then these variational algorithms employ an extra minimization step in order to approximate an optimal input state for the channel fidelity.

Similarly, we can estimate the fidelity of strategies by employing a sequence of parameterized circuits to function as a co-strategy and then minimize over them, in conjunction with any of the previous methods for estimating fidelity of states.
G. Estimating maximum output fidelity of channels

In this section, we show how a simple variation of Algorithm 8, in which we combine the actions of the min-prover and max-prover into a single max-prover, leads to a QIP algorithm for estimating the following fidelity function of two quantum channels $\mathcal{N}^0_{A\rightarrow B}$ and $\mathcal{N}^1_{A\rightarrow B}$:

$$F_{\text{max}}(\mathcal{N}^0, \mathcal{N}^1) := \sup_{\rho_A} F(\mathcal{N}^0_{A\rightarrow B}(\rho_A), \mathcal{N}^1_{A\rightarrow B}(\rho_A)),$$ (78)

where the optimization is over every input state $\rho_A$. This algorithm is based in part on Algorithm 4 but instead features an optimization over input states of the prover.

**Algorithm 10** The algorithm proceeds as follows:

1. The verifier prepares a Bell state
   $$|\Phi\rangle_{TT} := \frac{1}{\sqrt{2}}(|00\rangle_{TT} + |11\rangle_{TT})$$ (79)
   on registers $T'$ and $T$ and prepares system $E'$ in the all-zeros state $|0\rangle_{E'}$.

2. The prover transmits the system $A$ of the state $|\psi\rangle_{RA}$ to the verifier.

3. Using the circuits $U^0_{AE\rightarrow BE}$ and $U^1_{AE'\rightarrow BE'}$, the verifier performs the following controlled unitary:
   $$\sum_{i\in\{0,1\}} |i\rangle_{iT} \otimes U^i_{AE\rightarrow BE}.$$ (80)

4. The verifier transmits systems $T'$ and $E$ to the prover.

5. The prover prepares a system $F$ in the $|0\rangle_F$ state and acts on systems $T'$, $E$, and $F$ with a unitary $P_{T'E'\rightarrow T'F'}$ to produce the output systems $T''$ and $F''$, where $T''$ is a qubit system.

6. The prover sends system $T''$ to the verifier, who then performs a Bell measurement
   $$\{\Phi_{T''T}, I_{T''T} - \Phi_{T''T}\}$$ (81)
   on systems $T''$ and $T$. The verifier accepts if and only if the outcome $\Phi_{T''T}$ occurs.

Figure 11 depicts Algorithm 10.

**Theorem 5** The acceptance probability of Algorithm 10 is equal to

$$\frac{1}{2} \left(1 + \sqrt{F_{\text{max}}(\mathcal{N}^0_{A\rightarrow B}, \mathcal{N}^1_{A\rightarrow B})}\right).$$ (82)

**Proof.** The proof can be found in Appendix A 5. ■

H. Generalization to multiple states

In this section, we generalize Algorithm 4 to multiple states, by devising a quantum algorithm that tests how similar all the states of an ensemble are to each other. Suppose that we are given an ensemble $\{p(x), \rho^x_S\}_{x\in X}$ of states of system $S$, with $d = |X|$, and we would like to know how similar they are to each other. Then we can perform a test like that given in Algorithm 4, but it is a multiple-state similarity test. The main difference is
that the verifier prepares an initial entangled state that encodes the prior probabilities \( \{ p(x) \}_{x \in \mathcal{X}} \) and the algorithm employs \( d \)-dimensional control systems throughout, instead of qubit control systems. We suppose that, for all \( x \in \mathcal{X} \), there is a circuit \( U^{x}_{RS} \) that generates a purification \( |\psi^{x}\rangle_{RS} \) as follows:

\[
|\psi^{x}\rangle_{RS} := U^{x}_{RS}|0\rangle_{RS},
\]

\[
\rho^{x}_{S} = \text{Tr}_{R}[|\psi^{x}\rangle\langle\psi^{x}|_{RS}].
\]

**Algorithm 11** The algorithm proceeds as follows:

1. The verifier prepares a state

\[
|\Phi^{p}\rangle_{T'R} := \sum_{x \in \mathcal{X}} \sqrt{p(x)}|xx\rangle_{T'R}.
\]

on registers \( T' \) and \( T \) and prepares systems \( RS \) in the all-zeros state \( |0\rangle_{RS} \).

2. Using the circuits in the set \( \{ U^{x}_{RS} \}_{x \in \mathcal{X}} \), the verifier performs the following controlled unitary:

\[
\sum_{x \in \mathcal{X}} |x\rangle|x\rangle_{T} \otimes U^{x}_{RS}.
\]

3. The verifier transmits systems \( T' \) and \( R \) to the prover.

4. The prover prepares a system \( F \) in the \( |0\rangle_{F} \) state and acts on systems \( T' \), \( R \), and \( F \) with a unitary \( P_{T'RF \rightarrow T''F} \) to produce the output systems \( T'' \) and \( F' \), where \( T'' \) is a qudit system.

5. The prover sends system \( T'' \) to the verifier, who then performs a qudit Bell measurement

\[
\{ \Phi_{T''T}, I_{T''T} - \Phi_{T''T} \}
\]

on systems \( T'' \) and \( T \), where

\[
\Phi_{T''T} = |\Phi\rangle\langle\Phi|_{T''T},
\]

\[
|\Phi\rangle_{T''T} := \frac{1}{\sqrt{\sigma}} \sum_{x \in \mathcal{X}} |xx\rangle_{T''T}.
\]

The verifier accepts if and only if the outcome \( \Phi_{T''T} \) occurs.

**Theorem 6** The acceptance probability of Algorithm 11 is equal to

\[
p_{\text{adv}}(\{ p(x), \rho^{x}_{S} \}_{x \in \mathcal{X}}) := \frac{1}{d} \left[ \sup_{x \in \mathcal{X}} \sqrt{\sigma} \sqrt{F(\rho^{x}_{S}, \sigma_{S})} \right]^{2},
\]

where the optimization is over every density operator \( \sigma_{S} \). This acceptance probability is bounded from above by

\[
\frac{1}{d} + \frac{2}{d} \sum_{x,y \in \mathcal{X}: x \neq y} \sqrt{p(x)p(y)} \sqrt{F(\rho^{x}_{S}, \rho^{y}_{S})}.
\]

When \( d = 2 \), this upper bound is tight.

**Proof.** The proof can be found in Appendix A.6. □

**Corollary 7** The fact that the upper bound is achieved in Theorem 6 for \( d = 2 \) leads to the following identity for states \( \rho^{0}_{S} \) and \( \rho^{1}_{S} \) and probability \( p \in [0, 1] \):

\[
\left[ \sup_{\sigma_{S}} \sqrt{p} \sqrt{F(\rho^{0}_{S}, \sigma_{S})} + \sqrt{1-p} \sqrt{F(\rho^{1}_{S}, \sigma_{S})} \right]^{2} = 1 + 2 \sqrt{p(1-p)} \sqrt{F(\rho^{0}_{S}, \rho^{1}_{S})},
\]

where the optimization is over every density operator \( \sigma_{S} \).
The acceptance probability in (90) is proportional to the
secrecy measure discussed in [KRS09, Eq. (19)],
which is the same as the max-conditional entropy of
the following classical–quantum state:
\[
\sum_{x \in \mathcal{X}} p(x)|x\rangle\langle x| \otimes \rho^S_x.
\] (93)

Indeed, it is a measure of secrecy because if an eavesdropper
has access to system S and if \( \rho^S_x \approx \sigma \) for all \( x \in \mathcal{X} \)
and if \( p(x) \approx 1/d \), then it is difficult for the eavesdropper
to guess the classical message in system T (also, the fidelity
is close to one). According to [SDG+21, Remark 2.7] and
the expression in (A70) of Appendix A6, the acceptance probability
in (90) is also a measure of the symmetric
distinguishability of the classical–quantum state in (93),
and thus gives this measure an operational meaning.

The upper bound in (91) on the acceptance probability
has some conceptual similarity with known upper
bounds on the success probability in state discrimina-
tion [Mon08, Qiu08], in the sense that we employ the
bounds on the success probability in state discrimina-
tion with a circuit [Mon08, Qiu08], in the sense that we employ the
bounds on the success probability in state discrimina-
tion [Mon08, Qiu08], in the sense that we employ the
upper bound in (91) on the acceptance probability
has some conceptual similarity with known upper
bounds on the success probability in state discrimina-
tion [Mon08, Qiu08], in the sense that we employ the
bounds on the success probability in state discrimina-
tion [Mon08, Qiu08]. We note some similarities between the problem outlined here
and coherent channel discrimination considered recently in
[Wil20]. However, these two problems are ultimately
different in their objectives.

I. Generalization to multiple channels and
strategies

We now generalize Algorithms 8 and 11 to the case of
testing the similarity of an ensemble of channels. The
resulting algorithm thus has applications in the context
of private quantum reading [BDW18, DBW20], in which
one goal of such a protocol is to encode a classical mes-

}e into a channel selected randomly from an ensemble
of channels such that it is indiscernible by an eaves-
dropper who has access to the output of the channel.

We also remark at the end of this section about a generaliza-
tion of Algorithms 9 and 12 to the case of an ensemble
of n-turn quantum strategies.

Let us first consider the case of channels. In more
detail, let \( \{p(x), \mathcal{N}^x_{A \rightarrow B}\}_{x \in \mathcal{X}} \) be an ensemble of quantum
channels. Set \( d = |\mathcal{X}| \). We suppose that, for all \( x \in \mathcal{X} \),
there is a circuit \( U^x_{AE' \rightarrow BE} \) that generates an isometric
extension of the channel \( \mathcal{N}^x_{A \rightarrow B} \), in the following sense:
\[
\mathcal{N}^x_{A \rightarrow B}(\omega_A) = \operatorname{Tr}_E[U^x_{AE' \rightarrow BE}(\omega_A \otimes \langle 0|0\rangle_{E'}(U^x_{AE' \rightarrow BE})^\dagger)].
\] (94)
The following algorithm employs competing provers, simi-
lar to how Algorithm 8 does.

\textbf{Algorithm 12} The algorithm proceeds as follows:

1. The verifier prepares a state
\[
|\Phi^p\rangle_{T' \rightarrow T} := \sum_{x \in \mathcal{X}} \sqrt{p(x)}|x\rangle|x\rangle_{T' \rightarrow T}
\] (95)
on registers \( T' \) and \( T \) and prepares system \( E' \) in
the all-zeros state \( |0\rangle_{RS} \).

2. The min-prover transmits the system \( A \) of the state
\( |\psi\rangle_{RA} \) to the verifier.

3. Using the circuits in the set \( \{U^x_{AE' \rightarrow BE}\}_{x \in \mathcal{X}} \),
the verifier performs the following controlled unitary:
\[
\sum_{x \in \mathcal{X}} |x\rangle|x\rangle\langle x| \otimes U^x_{AE' \rightarrow BE}.
\] (96)

4. The verifier transmits systems \( T' \) and \( E \) to the
max-prover.

5. The max-prover prepares a system \( F \) in the \( |0\rangle_F \)
state and acts on systems \( T', R, \) and \( F \) with a uni-
\[\Phi^p \rightarrow \Phi^p \rightarrow \Phi^p\]
tary \( \Phi^p \rightarrow \Phi^p \rightarrow \Phi^p \) to produce the output systems \( F' \)
and \( F'' \), where \( F'' \) is a qudit system.

6. The max-prover sends system \( F'' \) to the verifier,
who then performs a qudit Bell measurement
\[
\{\Phi^p \rightarrow \Phi^p \rightarrow \Phi^p \} \rightarrow \Phi^p \rightarrow \Phi^p \rightarrow \Phi^p
\] (97)
on systems \( F'' \) and \( F' \), where \( \Phi^p \rightarrow \Phi^p \rightarrow \Phi^p \) is defined in (88). The verifier accepts if and only if the outcome
\[\Phi^p \rightarrow \Phi^p \rightarrow \Phi^p \] occurs.

\textbf{Theorem 8} The acceptance probability of Algorithm 12 is equal to
\[
p_{\text{Sim}}(\{p(x), \mathcal{N}^x_{A \rightarrow B}\}_{x \in \mathcal{X}}) =
\frac{1}{d} \left[ \inf_{\psi_{RA}} \sup_{\sigma_{RB}} \sum_{x \in \mathcal{X}} \sqrt{p(x)} \sqrt{\mathcal{F}(\mathcal{N}^x_{A \rightarrow B}(\psi_{RA}), \sigma_{RB})} \right]^2.
\] (98)

This acceptance probability is bounded from above by
\[
\frac{1}{d} + \frac{2}{d} \times \inf_{\psi_{RA}} \sum_{x, y \in \mathcal{X}} \sqrt{p(x)p(y)} \sup_{\sigma_{RB}} \mathcal{F}(\mathcal{N}^x_{A \rightarrow B}(\psi_{RA}), \mathcal{N}^y_{A \rightarrow B}(\psi_{RA}))
\] (99)

When \( d = 2 \), this upper bound is tight.

\textbf{Proof.} The proof can be found in Appendix A7. \( \blacksquare \)

\textbf{Corollary 9} The following identity holds in the special
case of two channels \( \mathcal{N}^0_{A \rightarrow B} \) and \( \mathcal{N}^1_{A \rightarrow B} \) and probability
\( p \in [0, 1] \):
\[
\left( \inf_{\psi_{RA}} \sup_{\sigma_{RB}} \left( \sqrt{p} \mathcal{F}(\mathcal{N}^0_{A \rightarrow B}(\psi_{RA}), \sigma_{RB}) \right) \right)^2
= 1 + \frac{2}{d} \left( \inf_{\psi_{RA}} \mathcal{F}(\mathcal{N}^0_{A \rightarrow B}(\psi_{RA}), \mathcal{N}^1_{A \rightarrow B}(\psi_{RA})) \right)^2.
\] (100)

where the supremum is with respect to every density op-
operator \( \sigma_{RB} \).
Remark 10 We note here that we can generalize the developments in this section and the previous one to the case of quantum strategies, in order to test how similar strategies in a set are to each other. Let \( \{p(x), N_{x}^{(n)}\}_{x \in X} \) be an ensemble of quantum strategies, each of which has \( n \) turns. Then the acceptance probability of an algorithm that is the obvious generalization of Algorithms 9 and 12 is given by

\[
\frac{1}{d} \left[ \inf_{S^{(n-1)}} \sup_{x \in X} \sqrt{p(x)} \cdot \sqrt{F\left(N_{x}^{(n)} \circ S^{(n-1)}, \sigma_{R_{n}B_{n}}\right)} \right]^2,
\]

where the infimum is with respect to every \((n-1)\)-turn pure co-strategy that leads to a quantum state \(N_{x}^{(n)} \circ S^{(n-1)}\) (as discussed around (66)) and the supremum is with respect to every state \(\sigma_{R_{n}B_{n}}\). The expression in (101) is a similarity measure for the strategies in the ensemble \(\{p(x), N_{x}^{(n)}\}_{x \in X}\).

We can also generalize Algorithm 10 from Section II G, to estimate the following similarity measure for an ensemble \(\{p(x), N_{A \rightarrow B}^{(n)}\}_{x \in X}\) of channels:

\[
\frac{1}{d} \left[ \sup_{\rho_{A}, \sigma_{B}} \sum_{x \in X} \sqrt{p(x)} \cdot \sqrt{F\left(N_{A \rightarrow B}^{x}(\rho_{A}), \sigma_{B}\right)} \right]^2,
\]

where the optimization is over all density operators \(\rho_{A}\) and \(\sigma_{B}\). As is the case with Algorithm 10, there is a single prover who is trying to make all of the channel outputs look like the same state. Again we suppose that there is a circuit \(U_{A \rightarrow E' \rightarrow B}^{T}E\) that generates an isometric extension of the channel \(N_{A \rightarrow B}^{x}\), in the sense of (94).

Algorithm 13 The algorithm proceeds as follows:

1. The verifier prepares a state

\[
|\Phi^p\rangle_{T' \rightarrow T} := \sum_{x \in X} \sqrt{p(x)} |xx\rangle_{T' \rightarrow T}
\]

on registers \(T'\) and \(T\) and prepares system \(E'\) in the all-zeros state \(|0\rangle_{E'}\).

2. The prover transmits the system \(A\) of the state \(|\psi\rangle_{RA}\) to the verifier.

3. Using the circuits in the set \(\{U_{A \rightarrow E' \rightarrow B}^{x}\}_{x \in X}\), the verifier performs the following controlled unitary:

\[
\sum_{x \in X} |x\rangle \langle x|_{T'} \otimes U_{A \rightarrow E' \rightarrow B}^{T}E.
\]

4. The verifier transmits systems \(T'\) and \(E\) to the max-prover.

5. The prover prepares a system \(F\) in the \(|0\rangle_{F}\) state and acts on systems \(T'\), \(R\), and \(F\) with a unitary \(P_{T'EF' \rightarrow T'F'}\) to produce the output systems \(T''\) and \(F''\), where \(T''\) is a qudit system.

6. The prover sends system \(T''\) to the verifier, who then performs a qudit Bell measurement

\[
\{\Phi_{T'' \rightarrow T}, I_{T'' \rightarrow T} - \Phi_{T'' \rightarrow T}\}
\]

on systems \(T''\) and \(T\), where \(\Phi_{T'' \rightarrow T}\) is defined in (88). The verifier accepts if and only if the outcome \(\Phi_{T'' \rightarrow T}\) occurs.

Theorem 11 The acceptance probability of Algorithm 13 is equal to

\[
p_{\sim}(\{p(x), N_{x}^{(n)}\}_{x \in X}) = \frac{1}{d} \left[ \sup_{\rho_{A}, \sigma_{B}} \sum_{x \in X} \sqrt{p(x)} \cdot \sqrt{F\left(N_{A \rightarrow B}^{x}(\rho_{A}), \sigma_{B}\right)} \right]^2.
\]

This acceptance probability is bounded from above by

\[
\frac{1}{d} + \frac{2}{d} \times \sup_{\rho_{A}, x,y \in X: x \neq y} \sum_{x \in X} \sqrt{p(x)p(y)} \cdot \sqrt{F\left(N_{A \rightarrow B}^{x}(\rho_{A}), N_{A \rightarrow B}^{y}(\rho_{A})\right)}.
\]

When \(d = 2\), this upper bound is tight.

Proof. For a fixed state \(\psi_{RA}\) of the prover, the problem is equivalent to that specified by Algorithm 11, for the ensemble \(\{p(x), F(N_{A \rightarrow B}^{x}(\rho_{A}))\}_{x \in X}\), where \(\rho_{A} = Tr_{A}[\psi_{RA}]\). Thus, all of the statements from Theorem 6 apply for this fixed state. We arrive at the statement of the theorem after optimizing over all input states.

III. ESTIMATING TRACE DISTANCE, DIAMOND DISTANCE, AND STRATEGY DISTANCE

We now review several well known algorithms for estimating trace distance [Wat02h], diamond distance [RW05], and strategy distance [GW07, Gut09, Gut12] by interacting with quantum provers. Later on, we replace the provers with parameterized circuits to see how well this approach can perform in estimating these distinguishability measures. A summary of the algorithms is presented in Table II.

A. Estimating trace distance

The trace distance between quantum states \(\rho_{S}^{A}\) and \(\rho_{S}^{B}\) is defined as \(\|\rho_{S}^{A} - \rho_{S}^{B}\|_{1}\), where \(\|A\|_{1} = Tr[\sqrt{A^{\dagger}A}]\). It is a well known and operationally motivated measure of distinguishability for quantum states.

We suppose, as is the case in Section II C, that quantum circuits \(U_{RS}^{i}\) are available for generating purifications of the states \(\rho_{S}^{A}\) and \(\rho_{S}^{B}\). That is, for \(i \in \{0,1\}\),

\[
\rho_{S}^{i} = Tr_{R}[U_{RS}^{i}|0\rangle_{RS}U_{RS}^{i\dagger}].
\]
- Generalizes Algorithm 14 to ensemble of states.

Algorithm 18 swaps the role of the max-prover and min-prover from Algorithm 17.

Algorithm 19

TABLE II. List of trace distance problems and algorithms addressed in this work. Approach used for each algorithm and comparison within a type of trace distance problem is also presented.

| Problem | Algorithms | Comparison |
|---------|------------|------------|
| $\| \rho_0 - \rho_1 \|_1$ | Algorithm 14 | Algorithm 14 does not require the purifying system, unlike fidelity algorithms. |
| $\| N_0 - N_1 \|_\infty$ | Algorithm 15 | - |
| $\| A^{n, (n)} - A^{r, (n)} \|_{\text{on}}$ | Algorithm 16 | - |
| $\| N_0 - N_1 \|_{\text{min}}$ | Algorithm 17 | - |
| $p_g(\{ p(x), \rho^x \})_{x \in X}$ | Algorithm 19 | Generalizes Algorithm 14 to ensemble of states. |

However, the purifying systems are not strictly necessary in the operation of the algorithm given below, which is an advantage over some of the algorithms from Section II C.

The following QSZK algorithm allows for estimating the trace distance [Wat02b], in the sense that its acceptance probability is a simple function of the trace distance:

**Algorithm 14 ([Hel67, Hel69, Hol72, Wat02b])**

*The algorithm proceeds as follows:*

1. The verifier picks a classical bit $i \in \{0,1\}$ uniformly at random, prepares the state $\rho^i_S$, and sends system $S$ to the prover.

2. The prover prepares a system $F$ in the $|0\rangle_F$ state and acts on systems $S$ and $F$ with a unitary $P_{SF{\rightarrow TF'}}$ to produce the output systems $T$ and $F'$, where $T$ is a qubit system.

3. The prover sends system $T$ to the verifier, who then performs a measurement on system $T$, with outcome $j \in \{0,1\}$. The verifier accepts if and only if $i = j$.

This algorithm has been well known for some time [Hel67, Hel69, Hol72, Wat02b] and its maximum acceptance probability is equal to

$$\max_{\Lambda: 0 \leq \Lambda \leq I} \frac{1}{2} \text{Tr}[\Lambda \rho^0_S] + \frac{1}{2} \text{Tr}[(I - \Lambda) \rho^1_S] = \frac{1}{2} \left( 1 + \frac{1}{2} \| \rho^0_S - \rho^1_S \|_1 \right).$$

(109)

This follows because the acceptance probability can be written as follows, for a fixed unitary $P = P_{SF{\rightarrow TF'}}$ of the prover:

$$\frac{1}{2} \sum_{i \in \{0,1\}} \text{Tr}[(i | i_T \otimes I_{F'}) P (\rho^i_S \otimes |0\rangle\langle 0_F|) P^\dagger]$$

$$= \frac{1}{2} \sum_{i \in \{0,1\}} \text{Tr}[(0 \otimes |0\rangle_F) P_{SF{\rightarrow TF'}} P^\dagger |0\rangle\langle 0_F| \rho^i_S]$$

(110)

$$= \frac{1}{2} \sum_{i \in \{0,1\}} \text{Tr}[\Lambda^i_S \rho^i_S],$$

(111)

where we have defined the measurement operator $\Lambda^i_S$, for $i \in \{0,1\}$, as

$$\Lambda^i_S := \langle 0_F | P_{SF{\rightarrow TF'}} P_{SF{\rightarrow TF'}}^\dagger |0\rangle_F \rho^i_S,$$

(112)

and it is clear that $\sum_{i \in \{0,1\}} \Lambda^i_S = I_S$. By the Naimark extension theorem [Nai40] (see also [KW20]), every measurement can be realized in this way, so that

$$\max_{\Lambda: 0 \leq \Lambda \leq I} \frac{1}{2} \text{Tr}[(i | i_T \otimes I_{F'}) P (\rho^i_S \otimes |0\rangle\langle 0_F|) P^\dagger]$$

$$= \max_{\Lambda: 0 \leq \Lambda \leq I} \frac{1}{2} \text{Tr}[\Lambda \rho^0_S] + \frac{1}{2} \text{Tr}[(I - \Lambda) \rho^1_S].$$

(113)

Thus, by replacing the actions of the prover with a parameterized circuit and repeating the algorithm, we can use a quantum computer to estimate a lower bound on the trace distance of the states $\rho^0_S$ and $\rho^1_S$. An approach similar to this has been adopted in [CSZW22].

We note here that the following identity holds also [Hel67, Hel69, Hol72] (see also [KW20, Theorem 3.13]):

$$\min_{\Lambda: 0 \leq \Lambda \leq I} \frac{1}{2} \text{Tr}[\Lambda \rho^0_S] + \frac{1}{2} \text{Tr}[(I - \Lambda) \rho^1_S]$$
\[
\frac{1}{2} \left( 1 - \frac{1}{2} \| \rho_S^{0} - \rho_S^{1} \|_1 \right). \tag{114}
\]

### B. Estimating diamond distance

The diamond distance between quantum channels \( N_A^{0} \rightarrow B \) and \( N_A^{1} \rightarrow B \) is defined as [Kit97]

\[
\left\| N_A^{0} \rightarrow B - N_A^{1} \rightarrow B \right\|_\diamond := \sup_{\rho_{RA}} \left\| N_A^{0} \rightarrow B(\rho_{RA}) - N_A^{1} \rightarrow B(\rho_{RA}) \right\|_1, \tag{115}\]

where the optimization is over every bipartite state \( \rho_{RA} \) and the system \( R \) can be arbitrarily large. By a well known data processing argument, the following equality holds

\[
\left\| N_A^{0} \rightarrow B - N_A^{1} \rightarrow B \right\|_\diamond := \max_{\psi_{RA}} \left\| N_A^{0} \rightarrow B(\psi_{RA}) - N_A^{1} \rightarrow B(\psi_{RA}) \right\|_1, \tag{116}\]

where the optimization is over every pure bipartite state \( \psi_{RA} \) and the system \( R \) is isomorphic to the channel input system \( A \). The diamond distance is a well known and operationally motivated measure of distinguishability for quantum channels [RW05, GLN05].

We suppose, as is the case in Section II D, that quantum circuits \( U_{AE}^{0} \rightarrow BE \) and \( U_{AE}^{1} \rightarrow BE \) are available for generating isometric extensions of the channels \( N_A^{0} \rightarrow B \) and \( N_A^{1} \rightarrow B \). That is, for \( i \in \{0, 1\} \),

\[
N_A^{i} \rightarrow B(\cdot) = \text{Tr}_E[U_{AE}^{i} \rightarrow BE(\cdot) \otimes |0\rangle \langle 0|]_{E'}. \tag{117}\]

However, the environment systems are not strictly necessary in the operation of the algorithm given below, which is an advantage over some of the algorithms from Section II D.

The following QIP algorithm allows for estimating the diamond distance [RW05], in the sense that its acceptance probability is a simple function of the diamond distance:

**Algorithm 15 ([RW05])** The algorithm proceeds as follows:

1. The prover prepares a pure state \( \psi_{RA} \) and sends system \( A \) to the verifier.
2. The verifier picks a classical bit \( i \in \{0, 1\} \) uniformly at random, applies the channel \( N_A^{i} \rightarrow B \), and sends system \( B \) to the prover.
3. The prover prepares a system \( F \) in the \( |0\rangle_F \) state and acts on systems \( R, B \), and \( F \) with a unitary \( P_{RBF} \rightarrow TF' \) to produce the output systems \( T \) and \( F' \), where \( T \) is a qubit system.
4. The prover sends system \( T \) to the verifier, who then performs a measurement on system \( T \), with outcome \( j \in \{0, 1\} \). The verifier accepts if and only if \( i = j \).
5. The prover applies the channel \( N_A^{1} \rightarrow B \) and sends system \( B \) to the verifier.

This algorithm has been well known for some time [RW05] and its maximum acceptance probability is equal to

\[
\frac{1}{2} \left( 1 + \frac{1}{2} \left\| N_A^{0} \rightarrow B - N_A^{1} \rightarrow B \right\|_\diamond \right). \tag{118}\]

Thus, by replacing the actions of the prover with a parameterized circuit and repeating the algorithm, we can use a quantum computer to estimate a lower bound on the diamond distance of the channels \( N_A^{0} \rightarrow B \) and \( N_A^{1} \rightarrow B \).

### C. Estimating strategy distance

We already provided the definition of a quantum strategy in Section II E, and therein, we discussed the strategy distance [GW07, CDP08, Gut12] is conceptually similar, but it is defined with the trace distance as the underlying metric:

\[
\left\| N_A^{0}(n) - N_A^{1}(n) \right\|_\diamond := \sup_{S^{(n-1)}} \left\| N_A^{0}(n) \circ S^{(n-1)} - N_A^{1}(n) \circ S^{(n-1)} \right\|_1, \tag{119}\]

where the supremum is with respect to every co-strategy \( S^{(n-1)} \) that leads to the quantum states \( N_A^{0}(n) \circ S^{(n-1)} \) and \( N_A^{1}(n) \circ S^{(n-1)} \) (here we have employed the same notation used in (66)). The strategy distance is an operationally motivated measure of distinguishability for quantum strategies.

The following QIP algorithm allows for estimating the strategy distance [GW07], in the sense that its acceptance probability is a simple function of the strategy distance:

**Algorithm 16 ([GW07])** The algorithm proceeds as follows:

1. The prover prepares a pure state \( \psi_{RA} \) and sends system \( A \) to the verifier.
2. The verifier picks a classical bit \( i \in \{0, 1\} \) uniformly at random, applies the channel \( N_A^{i} \rightarrow M_{1}B_{1} \), and sends system \( B_{1} \) to the prover.
3. The prover acts with the isometric channel \( S_{R_{1}B_{1} \rightarrow R_{2}A_{2}}^{i} \) and then sends system \( A_{2} \) to the verifier.
4. For \( k \in \{2, \ldots, n-1\} \), the verifier applies the channel \( N_{A_{k-1}}^{i} \rightarrow M_{k}B_{k} \) and transmits system \( B_{k} \) to the prover, who subsequently acts with the isometric channel \( S_{R_{k}B_{k} \rightarrow R_{k+1}A_{k+1}}^{i} \) and then sends system \( A_{k+1} \) to the verifier.
5. The verifier applies the channel \( N_{M_{n-1}A_{n}}^{i,n} \rightarrow B_{n} \) and sends system \( B_{n} \) to the prover.
6. The prover prepares a system $F$ in the $|0\rangle_F$ state and acts on systems $R_n$, $B_n$, and $F$ with a unitary $P_{R_nB_nF\rightarrow TF'}$ to produce the output systems $T$ and $F'$, where $T$ is a qubit system.

7. The prover sends system $T$ to the verifier, who then performs a measurement on system $T$, with outcome $j \in \{0, 1\}$. The verifier accepts if and only if $i = j$.

This algorithm has been well known since [GW07] and its maximum acceptance probability is equal to

$$\frac{1}{2} \left( 1 + \frac{1}{2} \left\| \mathcal{N}^{\|\cdot\|_1}_{\text{on}} - \mathcal{N}^{\|\cdot\|_1}_{\text{on}} \right\| \right).$$

Thus, by replacing the actions of the prover with a parameterized circuit and repeating the algorithm, we can use a quantum computer to estimate a lower bound on the strategy distance of the strategies $\mathcal{N}^{\|\cdot\|_1}_{\text{on}}$ and $\mathcal{N}^{\|\cdot\|_1}_{\text{on}}$. See [Gut12, KW21a] for semi-definite programs for evaluating the strategy distance of two strategies.

D. Estimating minimum trace distance of channels

In this section, we show how to estimate the following trace distance function of channels $\mathcal{N}^{\|\cdot\|_1}_{AB}$ and $\mathcal{N}^{\|\cdot\|_1}_{AB}$ by means of a short quantum game (SQG) algorithm:

$$\inf_{\rho_A} \left\| \mathcal{N}^{\|\cdot\|_1}_{AB} - \mathcal{N}^{\|\cdot\|_1}_{AB} \right\|_1,$$

where the optimization is over every input state $\rho_A$. The algorithm features a min-prover and a max-prover. Short quantum games were defined and studied in [GW05, Gut05].

Algorithm 17 The algorithm proceeds as follows:

1. The min-prover prepares a state $\psi_{RA}$ and sends system $A$ to the verifier.

2. The verifier picks a classical bit $i \in \{0, 1\}$ uniformly at random, applies the channel $\mathcal{N}_{A\rightarrow B}$, and sends system $B$ to the max-prover.

3. The max-prover prepares a system $F$ in the $|0\rangle_F$ state and acts on systems $R$, $B$, and $F$ with a unitary $P_{RBF\rightarrow TF'}$ to produce the output systems $T$ and $F'$, where $T$ is a qubit system.

4. The max-prover sends system $T$ to the verifier, who then performs a measurement on system $T$, with outcome $j \in \{0, 1\}$. The verifier accepts if and only if $i = j$.

For a fixed state $\psi_{RA}$ of the max-prover, it follows from (114) that the acceptance probability is equal to

$$\frac{1}{2} \left( 1 + \frac{1}{2} \left\| \mathcal{N}^{\|\cdot\|_1}_{AB} - \mathcal{N}^{\|\cdot\|_1}_{AB} \right\|_1 \right),$$

where $\rho_A = \text{Tr}_R[\psi_{RA}]$. Since the min-prover plays first and his goal is to minimize the acceptance probability, it follows that the acceptance probability of Algorithm 17 is given by

$$\frac{1}{2} \left( 1 + \frac{1}{2} \left\| \mathcal{N}^{\|\cdot\|_1}_{AB} - \mathcal{N}^{\|\cdot\|_1}_{AB} \right\|_1 \right).$$

Another way to estimate the minimum trace distance of channels in (121) is to swap the roles of the max-prover and min-prover in Algorithm 17:

Algorithm 18 The algorithm proceeds as follows:

1. The max-prover prepares a state $\psi_{RA}$ and sends system $A$ to the verifier.

2. The verifier picks a classical bit $i \in \{0, 1\}$ uniformly at random, applies the channel $\mathcal{N}_{A\rightarrow B}$, and sends system $B$ to the min-prover.

3. The min-prover prepares a system $F$ in the $|0\rangle_F$ state and acts on systems $R$, $B$, and $F$ with a unitary $P_{RBF\rightarrow TF'}$ to produce the output systems $T$ and $F'$, where $T$ is a qubit system.

4. The min-prover sends system $T$ to the verifier, who then performs a measurement on system $T$, with outcome $j \in \{0, 1\}$. The verifier accepts if and only if $i = j$.

For a fixed state $\psi_{RA}$ of the max-prover, it follows from (114) that the acceptance probability is equal to

$$\frac{1}{2} \left( 1 + \frac{1}{2} \left\| \mathcal{N}^{\|\cdot\|_1}_{AB} - \mathcal{N}^{\|\cdot\|_1}_{AB} \right\|_1 \right),$$

where $\rho_A = \text{Tr}_R[\psi_{RA}]$. Since the max-prover plays first and his goal is to maximize the acceptance probability, it follows that the acceptance probability of Algorithm 17 is given by

$$\frac{1}{2} \left( 1 - \frac{1}{2} \inf_{\rho_A} \left\| \mathcal{N}^{\|\cdot\|_1}_{AB} - \mathcal{N}^{\|\cdot\|_1}_{AB} \right\|_1 \right).$$

Although the quantities estimated by Algorithms 10 and 17 or 18 are similar (and related to each other by standard inequalities relating trace distance and fidelity [FvdG99]), the algorithms are very different in that the channel output is available at the end of Algorithm 10, whereas it is not at the end of Algorithms 17 and 18. This has implications for applications in which it is helpful to have access to the channel output, for example, when one is trying to find the fixed point of a quantum channel.
E. Generalization to multiple states, channels, and strategies

Each of the algorithms from the previous subsections has a generalization to multiple states, channels, and strategies. We go through them briefly here. The main idea is that, rather than randomly picking from a set of two resources, the verifier picks randomly from a set of multiple resources and then a prover has to guess which one was chosen. The main difference with the binary case is that there is not a closed-form expression for the acceptance probability in terms of a metric like the trace distance or derived metrics, but rather the optimization is phrased as a semi-definite program that can be solved numerically or used in some cases to obtain analytical solutions (for example, if there is sufficient symmetry).

Suppose that we are given an ensemble \( \{ p(x), \rho^x_S \}_{x \in \mathcal{X}} \) of quantum states. The verifier picks \( x \) randomly according to \( p(x) \), prepares \( \rho^x_S \), and the prover has to guess which state was prepared. The acceptance probability is given by

\[
p_y(\{ p(x), \rho^x \}_{x \in \mathcal{X}}) := \sup_{\{ \Lambda^x_S \}_{x \in \mathcal{X}}} \sum_{x \in \mathcal{X}} p(x) \Tr[\Lambda^x_S \rho^x_S],
\]

(127)

where the optimization is over every POVM \( \{ \Lambda^x_S \}_{x \in \mathcal{X}} \). In the case that \( |\mathcal{X}| = 2 \), this acceptance probability has the explicit form

\[
\frac{1}{2} \left( 1 + \| p \rho^0_S - (1 - p) \rho^1_S \|_1 \right).
\]

(128)

To account for multiple states, we modify Algorithm 14 as follows: the verifier’s variable \( i \in \{ 0, \ldots, |\mathcal{X}| - 1 \} \) is randomly selected and the prover’s guess \( j \) is chosen from the same set. System \( T \) therein is generalized to be a \( \lceil \log_2 |\mathcal{X}| \rceil \)-qubit system. When \( |\mathcal{X}| \) is a power of two, there is a perfect match between the number \( |\mathcal{X}| \) of measurement outcomes and the dimension of system \( T \). The verifier accepts if the outcome \( j \) equals the state \( i \) that was picked. If \( |\mathcal{X}| \) is not a power of two, the following algorithm handles this case by coarse graining some of the measurement outcomes together. This is relevant because most quantum computers are qubit-based.

**Algorithm 19** The algorithm proceeds as follows:

1. The verifier selects an integer \( i \in \{ 0, \ldots, |\mathcal{X}| - 1 \} \) at random according to \( p(i) \), prepares the state \( \rho^i_S \), and sends system \( S \) to the prover.

2. The prover prepares a system \( F \) composed of \( \lceil \log_2 |\mathcal{X}| \rceil \) qubits in the \( |0\rangle_F \) state. The prover then acts on systems \( S \) and \( F \) with a unitary \( P_{SF \rightarrow TF'} \), producing the output systems \( F' \) and \( T \), where \( T \) is a system of \( \lceil \log_2 |\mathcal{X}| \rceil \) qubits.

3. The prover sends system \( T \) to the verifier, who then performs a computational basis measurement on system \( T \), with outcome \( j \in \{ 0, \ldots, 2^{\lceil \log_2 |\mathcal{X}| \rceil} - 1 \} \).

4. The verifier accepts under two conditions.

   - \( j \leq |\mathcal{X}| - 1 \) and \( i = j \).
   - \( j > |\mathcal{X}| - 1 \) and \( i = 0 \).

This algorithm is a direct generalization of Algorithm 14. To understand its connection to (127), consider that, for a fixed unitary \( P_{SF \rightarrow TF'} \), its acceptance probability is given by

\[
\sum_{i \in \{ 0, \ldots, |\mathcal{X}| - 1 \}} p(i) \Tr[\rho^i_S \otimes |0\rangle \langle 0|_F P_1] + p(0) \sum_{j = |\mathcal{X}|} \Tr[\rho^j_S \otimes |0\rangle \langle 0|_F P_1] \]

(129)

\[
= \sum_{i \in \{ 0, \ldots, |\mathcal{X}| - 1 \}} p(i) \Tr[|0\rangle_F P_1(\rho^i_S |i\rangle \langle i|_T \otimes I_{T'}) P_0 |0\rangle_F + \rho^j_S] \]

(130)

\[
+ p(0) \sum_{j = |\mathcal{X}|} \Tr[|0\rangle_F P_1(\rho^j_S |j\rangle \langle j|_T \otimes I_{T'}) P_0 |0\rangle_F + \rho^j_S].
\]

(131)

where we have defined the following measurement operators:

\[
\Lambda^i_S := \langle 0\rangle_F P_1(\rho^i_S |i\rangle \langle i|_T \otimes I_{T'}) P_0 |0\rangle_F \]

(132)

\[
+ \sum_{j = |\mathcal{X}|} \langle 0\rangle_F P_1(\rho^j_S |j\rangle \langle j|_T \otimes I_{T'}) P_0 |0\rangle_F,
\]

and for all \( i \in \{ 0, 1, \ldots, |\mathcal{X}| - 1 \} \):

\[
\Lambda^i_S := \langle 0\rangle_F P_1(\rho^i_S |i\rangle \langle i|_T \otimes I_{T'}) P_0 |0\rangle_F.
\]

(133)

As such, we coarsen all measurement outcomes in \( \{ 0, |\mathcal{X}|, |\mathcal{X}| + 1, \ldots, 2^{\lceil \log_2 |\mathcal{X}| \rceil} - 1 \} \) into a single measurement outcome. By the Naimark extension theorem, every measurement with \( |\mathcal{X}| \) outcomes can be realized in this way, so that maximizing the expression in (129) over every unitary \( P \) gives a value equal to that in (127).

On the one hand, if \( |\mathcal{X}| \) is a power of two, then it follows that \( |\mathcal{X}| = 2^{\lceil \log_2 |\mathcal{X}| \rceil} \) and the outcome \( j > |\mathcal{X}| - 1 \) never occurs. On the other hand, if \( |\mathcal{X}| \) is not a power of two, then \( |\mathcal{X}| < 2^{\lceil \log_2 |\mathcal{X}| \rceil} \) and the outcome \( j > |\mathcal{X}| - 1 \) does occur.

Now suppose that we are given an ensemble \( \{ p(x), \mathcal{N}^x \}_{x \in \mathcal{X}} \) of quantum channels. Then a similar modification of Algorithm 15 has acceptance probability

\[
\sup_{\psi_{RA}, \{ \Lambda^x_{RB} \}_{x \in \mathcal{X}}} \sum_{x \in \mathcal{X}} p(x) \Tr[\Lambda^x_{RB} \mathcal{N}^x_{A \rightarrow B}(\psi_{RA})],
\]

(134)

where the optimization is over every state \( \psi_{RA} \) and POVM \( \{ \Lambda^x_{RB} \}_{x \in \mathcal{X}} \). In the case that \( |\mathcal{X}| = 2 \), this acceptance probability has the explicit form

\[
\frac{1}{2} \left( 1 + \| p \mathcal{N}^0_{A \rightarrow B} - (1 - p) \mathcal{N}^1_{A \rightarrow B} \|_\infty \right).
\]

(135)
Suppose we are given an ensemble \( \{ p(x), N^{x,n}(x) \}_{x \in X} \) of \( n \)-turn quantum strategies. A similar modification of Algorithm 16 has acceptance probability

\[
\sup_{S^{(n-1)}, (A_B^{(n+1)})_{x \in X}} \sum_{x \in X} p(x) \Tr[N^{x,n}(x) \circ S^{(n-1)}],
\]

where the optimization is over every \((n-1)\)-turn pure co-strategy \( S^{(n-1)} \) and POVM \( \{ A_B^{(n+1)} \}_{x \in X} \). In the case that \(|X| = 2\), this acceptance probability has the explicit form

\[
\frac{1}{2} \left( 1 + \| pN^{0,n}(n) - (1-p)N^{1,n}(n) \|_\infty \right),
\]

where this is the strategy norm.

Finally, we can generalize Algorithms 17 and 18, with the acceptance probabilities respectively given by

\[
\inf_{\rho_A} \sup_{\{ A_B^{(n+1)} \}_{x \in X}} \sum_{x \in X} p(x) \Tr[\Lambda^{x,n}_B \circ \rho_A A_B^{(n+1)}],
\]

\[
\sup_{\rho_A} \inf \sum_{x \in X} p(x) \Tr[\Lambda^{x,n}_B \circ \rho_A A_B^{(n+1)}].
\]

In the case that \(|X| = 2\), these acceptance probabilities become

\[
\frac{1}{2} \left( 1 + \inf_{\rho_A} \| pN^{0,n}_A \circ B - (1-p)N^{1,n}_A \circ B \|_1 \right),
\]

\[
\frac{1}{2} \left( 1 - \inf_{\rho_A} \| pN^{0,n}_A \circ B - (1-p)N^{1,n}_A \circ B \|_1 \right).
\]

IV. PERFORMANCE EVALUATION OF ALGORITHMS USING A NOISY AND NOISY QUANTUM SIMULATOR

In this section, we present results obtained from numerically simulating Algorithms 4–7 and Algorithm 14 on a noiseless quantum simulator and Algorithms 8, 15, and 19 on both a noiseless and noisy quantum simulator. In the first subsection, we introduce and discuss the circuit ansatz employed in these numerical experiments. In the next subsection, we discuss the form of the states and channels used for the numerical simulations. In the following subsections, we present the details of our numerical simulations of Algorithms 4–7 for fidelity of states, Algorithm 8 for the fidelity of channels, Algorithm 14 for trace distance of states, Algorithm 15 for diamond distance of channels, and Algorithm 19 for multiple state discrimination.

In the simulations below, we use a maximum number of iterations to be the stopping condition. We noted that some algorithms - in particular, ones with multiple provers - were more prone to get stuck in local minima and optimization loops. We found that, in these scenarios, using convergence as the stopping condition could lead to an unbounded number of iterations. In these cases, we found that using a maximum number of iterations was sufficient and effective.

All the program code for Algorithms 4, 5, 6, 7, 8, 14, 15, 19, and corresponding SDPs can be found as arXiv ancillary files with the arXiv posting of this paper.

A. Ansatz

To estimate the relevant quantities in this work, we employ the hardware-efficient ansatz (HEA) \([KMT+17]\). The HEA is a problem-agnostic ansatz that depends on the architecture and the connectivity of the given hardware. In this work, we consider a fixed structure of the HEA. Let \( X, Y, \) and \( Z \) denote the Pauli matrices. We define one layer of the HEA to consist of single-qubit rotations \( e^{-i\theta/2} e^{-i\delta/2} \), each of which acts on a single qubit and is parameterized by \( \theta \) and \( \delta \), followed by CNOTs between neighboring qubits. A CNOT between the control qubit \( k \) and the target qubit \( \ell \) is given by

\[
e^{-i\pi/2(|1\rangle_k \otimes \langle X|_{\ell})} = |0\rangle_k \otimes I_\ell + |1\rangle_k \otimes X_\ell. \]

For our numerical experiments, we consider a sufficiently large number of layers of the HEA. In principle, both the circuit structure and the number of layers of the HEA can be made random and this randomness can lead to better performance of variational algorithms \([BCV+21]\). We leave the study of such ansatze for future work.

The HEA is used both to create the states and channels, as well as to create a parameterized unitary that replaces the provers. In the former two cases, the rotation angles are fixed, but in the prove scenario, the angles are parameters that are optimized.

B. Test states and channels

To study the performance of our algorithms, we randomly select states and channels as follows. For \( n \)-qubit states, we apply \( m \) layers of the HEA with randomly selected angles for rotation around the \( x \)- and \( y \)-axes on \( n + k \) qubits initialized to the state \(|0\rangle_k \otimes |0\rangle_0\). This procedure prepares a pure state on \( n + k \) qubits and hence, a mixed state on \( n \) qubits of rank \( \leq 2^k \).

To realize an \( n \)-qubit channel \( N_{A\to B} \), we generate a unitary \( U_{A\to E'} \) on \( n + k \) qubits such that

\[
N_{A\to B}(\omega_A) := \Tr_E [U_{A\to E'} \otimes |0\rangle_{E'} \langle 0| E') (U_{A\to E'} \otimes |0\rangle_{E'}),
\]

where systems \( E' \) and \( E \) each consist of \( k \) qubits. Due the Stinespring dilation theorem \([Sti55]\), this is a general approach by which arbitrary channels can be realized.
For our experiments, we set $U$ to consist of $m$ layers of the HEA itself, with randomly selected angles for rotation around the $x$- and $y$-axes on $n + 1$ qubits. Tracing out one of the qubits gives a channel on $n$ qubits, as required.

Several algorithms in our paper (see (6), (20), (60)) depend on having access to unitaries of the form

$$
\sum_{i \in \{0, 1\}} |i\rangle_i \otimes U_S^0 = |0\rangle_0 \otimes U_S^0 + |1\rangle_1 \otimes U_S^1. \quad (144)
$$

These can be split into the sequential application of the following two controlled unitaries:

$$
|0\rangle_0 \otimes I + |1\rangle_1 \otimes U^1_S, \\
|1\rangle_1 \otimes I + |0\rangle_0 \otimes U^0_S, \quad (145)
$$

of which our algorithms make use.

### C. Fidelity of states

In this section, we discuss the performance of Algorithms 4–7 in the noiseless scenario to estimate the fidelity between two three-qubit mixed states. Algorithms 4–7 require different numbers of qubits for estimating the fidelity between $\rho$ and $\sigma$. In particular, for this case, Algorithm 4 requires eight qubits, along with access to controlled unitaries, as defined in (145). Algorithms 5, 6, and 7 require 13, 10, and 8 qubits, respectively. We recall that Algorithms 4–6 require purifications of both $\rho$ and $\sigma$, while Algorithm 7 relies only on access to $\rho$ and $\sigma$ directly. Moreover, Algorithms 4 and 5 require measurements on two qubits, and Algorithm 6 requires Bell measurements on ten qubits. Finally, Algorithm 7 requires two single-qubit measurements.

We now summarize the HEA employed. For Algorithm 4, the prover unitary is created using five layers of the HEA, which acts on four qubits. Similarly, in Algorithm 5, we employ eight layers of the HEA that acts on six qubits. In Algorithm 6, the ansatz acts on two qubits, and we consider four layers of it. In Algorithm 7, the ansatz acts on four qubits, and we apply eight layers of it. For our implementations, we picked these circuit depths so that the cost function is minimized. A more general framework allows for the ansatz structure to be unfixed and instead variable, but we leave the detailed study of this, for our algorithms, to future work [BCV+21].

We begin the training with a random set of variational parameters. We evaluate the cost using a state vector simulator (noiseless simulator) [AAMA+21]. We then employ the gradient-descent algorithm to obtain a new set of parameters. We note that in general, the true fidelity between states $\rho$ and $\sigma$ is not known. Thus the stopping criterion for these algorithms is a maximum number of iterations. For our numerical experiments, we set the total number of iterations to be 300. For each algorithm, we run ten instances of the algorithm and pick the best run for generating Figure 12.

### FIG. 12. Estimation of the fidelity between quantum states versus the number of iterations. We implement Algorithms 4–7 on a noiseless simulator to estimate the fidelity between two three-qubit mixed states, each of rank $\leq 4$. For each variational algorithm, we employ the HEA, as defined in Section IV.A. In particular, we start with a random parameter vector $\vec{\theta}$ and then update it according to a gradient-based optimization procedure. The dashed-dotted line represents the true fidelity between two randomly chosen quantum states. In each case, the optimization procedure converges to the true fidelity with high accuracy. Algorithms 4–7 achieve an absolute error in fidelity estimation of order $10^{-5}, 10^{-4}, 10^{-9}$, and $10^{-3}$, respectively.

In Figure 12, we plot the results of the numerical simulations. The dashed-dotted line represents the true fidelity between two random three-qubit quantum states $\rho$ and $\sigma$, as described above. Each algorithm converges to the true fidelity with high accuracy within a finite number of iterations. As discussed above, for each algorithm, the HEA is of a different size. Thus, it is not straightforward to compare these different algorithms. In terms of the convergence rate, we find that Algorithm 6 converges to the true fidelity faster than all other algorithms. Algorithms 4–7 achieve an absolute error in fidelity estimation of order $10^{-5}, 10^{-4}, 10^{-9}$, and $10^{-3}$, respectively.

### D. Trace distance of states

Using Algorithm 14, we estimate the normalized trace distance $\frac{1}{2} \| \rho - \sigma \|$ between two three-qubit states $\rho$ and $\sigma$, each having rank $\leq 4$, as defined above in Section IV.B. For our numerical experiments, we use a noiseless simulator. Algorithm 14 requires eight qubits in total and two single-qubit measurements. We employ ten layers of the HEA, which acts on four qubits. Similar to the fidelity-estimation algorithms detailed above, we begin with a random set of variational parameters and update them using the gradient-descent algorithm.

As the true normalized trace distance between $\rho$ and $\sigma$
is assumed to be unknown, we use a stopping criterion as the number of iterations, which we take to be 300 iterations. For Algorithm 14, we run ten instances of it and pick the best run for generating Figure 13.

In Figure 13, we plot the results of Algorithm 14. The dashed-dotted line represents the true normalized trace distance between two random three-qubit quantum states $\rho$ and $\sigma$, as described above. The absolute error in trace-distance estimation is of order $10^{-4}$.

### E. Fidelity of channels

In this section, we discuss the performance of Algorithm 8 in both the noiseless and noisy scenarios. The channels in question are realized by using parameterized unitaries and tracing out ancilla qubits, as discussed in Section IV B. The algorithm employs a min-max optimization and thus requires two parameterized unitaries representing the min- and max-provers, respectively. The controlled unitaries consist of one layer of the HEA, with each consisting of random rotations about the $x$-axis, on two qubits, thereby realizing the $\mathcal{N}_{i}^{A\rightarrow B}$ channels acting on one qubit, for $i \in \{0, 1\}$.

We now summarize the HEA employed in generating the min- and max-provers. The min-prover unitary is generated using two layers of the HEA, which acts on two qubits. The max-prover unitary is generated using two layers of the HEA, which acts on three qubits. The rotation angles for both provers around the $x$- and $y$-axes are chosen at random. The particular choices of the number of layers are made so that the cost function is minimized.

We begin the training phase with a random set of variational parameters for both parameterized unitaries. For the noiseless simulation, we evaluate the cost using a state vector simulator (noiseless simulator) [AAMA+21]. For the noisy simulation, we use the QASM simulator with the noise model from IBM-Jakarta. Since the number of parameters is significantly higher than the previous algorithms, to speed up the convergence, we employ both the simultaneous perturbation stochastic approximation (SPSA) method [Spa98] and the gradient-descent method to obtain a new set of parameters.

The optimization is carried out in a zig-zag fashion, explained as follows. The minimizing optimizer implements the SPSA algorithm and is allowed to run until convergence occurs. Then, the maximizing optimizer, implementing the gradient descent algorithm, runs for one iteration. We note that in general, the true fidelity between the channels $\mathcal{N}^{0}$ and $\mathcal{N}^{1}$ is not known. Thus, the stopping criterion for these algorithms is a maximum number of iterations. For our numerical experiments, we set the total number of iterations to be 6000, mostly used in the minimizing optimizer. The results of the numerical simulations are presented in Figure 14.

Note that the graph presented in Figure 14 shows that the convergence is highly non-monotonic, unlike the convergence behavior presented in previous graphs. Each iteration consists of a decrease in the function value, followed by a single increasing iteration. This is clearly indicative of the min-max optimization nature of the algorithm. Furthermore, unlike other algorithms, the optimization value in this algorithm can overshoot the true solution, due to the min-max nature of the optimization. However, the noiseless plot indicates that, once it overshoots the solution, it oscillates with decreasing amplitude and converges.

The noisy optimization converges as well, but it does not converge to the known value of the root fidelity of the two channels. However, the parameters found after convergence exhibit a noise resilience, as put forward in [SKCC20]; i.e., using the parameters obtained from the noisy optimization in a noiseless simulator gives a value much closer to the true value, as indicated by the solid orange line in Figure 14.

### F. Diamond distance of channels

In this section, we discuss the performance of Algorithm 15 in the noiseless and noisy scenarios. Algorithm 15 requires eight qubits. Similar to the previous section, the channels in question are realized using the procedure from Section IV B. The algorithm utilizes a max-max optimization and thus requires two parameterized unitaries representing the two max-provers. Each unitary $U_{A_i \rightarrow B_i}^{i}$, for $i \in \{0, 1\}$, consists of one layer of the HEA with random rotations about the $x$- and $y$-axes, on two qubits, each thereby realizing the one-qubit channel $\mathcal{N}_{i}^{A_i \rightarrow B_i}$.
We now summarize the HEA employed in generating the two provers. The first prover, called the state-prover because its goal is to realize an optimal distinguishing state, is generated using two layers of the HEA, which acts on two qubits. The second prover, called the max-prover, is generated using two layers of the HEA, which acts on three qubits. The rotation angles for both provers around the $x$- and $y$-axes are chosen at random. The particular choices of the number of layers are made so that the cost function is minimized.

We begin the training phase with a random set of variational parameters for both parameterized unitaries. In the noiseless simulation, we evaluate the cost using a state vector simulator (noiseless simulator). In the noisy setup, we use the QASM-simulator with the noise model from IBM-Jakarta. Similar to the previous section, we employ the SPSA optimization technique.

The optimization is carried out in two parts—the first part uses the COBYLA optimizer [Pow94, VGO+20] (non-gradient based), and the second part uses the SPSA optimizer. In both stages, the optimization is carried out in a zig-zag fashion, explained as follows. The first stage allows for moving quickly into the neighbourhood of the actual solution, but then slows down dramatically. Once we approach the solution, we switch to a gradient-based method that converges to the solution more quickly. In both stages, we allow the state-prover and the max-prover to be optimized for a fixed number of iterations in a zig-zag manner. This is because, in general, the true diamond distance between channels $\mathcal{N}^0$ and $\mathcal{N}^1$ is not known. Thus the stopping criterion for these algorithms is a maximum number of iterations. For our numerical experiments, we set the total number of iterations to be 1600. The results of the numerical simulations are presented in Figure 15.

Note that the noiseless graph presented in Figure 15 shows that the convergence is highly monotonic, unlike the fidelity of channels (see Figure 14), because the optimization is a max-max one, as opposed to the min-max nature of Algorithm 8. The quick convergence, indicated by the lower number of iterations, is a consequence of this difference.

The noisy simulation converges as well, and similar to the previous section, the parameters exhibit a noise resilience. Once the COBYLA stage of the optimization is completed, the SPSA optimization is more noisy, due to the perturbative nature of the algorithm. Note that the COBYLA optimizer operates in batches of 30, giving an impression of smoothness.

G. Multiple state discrimination

In this section, we discuss the performance of Algorithm 19 in the noisy and noiseless scenarios. We consider a specific scenario of distinguishing three one-qubit mixed states. Recall from Section IV B that the one-qubit states are generated by using two layers of the HEA on two qubits. We execute this on a qubit system, and hence we use Algorithm 19. The algorithm requires twelve qubits in total and three two-qubit measurements. The measurement is realized using a parameterized unitary and ancilla qubits. By Naimark’s extension theorem [Nai40],
our algorithms with quantum computational complexity theory. In this section, we prove that several basic quantum complexity classes can be reframed as distance and fidelity estimation problems. That is, we show that various distance and fidelity estimation problems are complete for various quantum complexity classes. Refs. [Wat09a, VW16] provide reviews of basic concepts in quantum computational complexity theory for interested readers.

In particular, here we summarize existing results linking estimation problems to complexity classes, and furthermore, we prove that five new distance estimation algorithms that are complete for some complexity classes of interest. First, we prove that promise versions of the following estimation problems are BQP-complete:

1. estimating the fidelity between two pure states,
2. estimating the fidelity between a pure state and a mixed state,
3. estimating the Hilbert–Schmidt distance of two arbitrary states.

Fourth, we prove that the promise problem version of estimating the fidelity between a pure state and a channel with arbitrary input is QMA-complete. Finally, we show that the promise problem version of estimating the fidelity between a pure state and a channel with a separable input state is QMA(2)-complete. In Figure 17, we summarize the various quantum complexity classes and the representative fidelity and distance estimation algorithms.

V. ESTIMATING DISTANCE MEASURES AS COMPLEXITY CLASSES

We now turn our attention to the intersection of our algorithms with quantum computational complexity theory. In this section, we prove that several basic quantum complexity classes can be reframed as distance and fidelity estimation problems. That is, we show that various distance and fidelity estimation problems are complete for various quantum complexity classes.

1. Completeness: For all \( x \in A_{\text{yes}} \),
\[
\Pr[Q \text{ accepts } x]
\]
is BQP-complete.

Theorem 12

The promise problem Fidelity-Pure-Pure is BQP-complete.

1. $(\alpha, \beta)$-Fidelity-Pure-Pure is in BQP for all $\alpha < \beta$. (It is implicit that the gap between $\alpha$ and $\beta$ is larger than an inverse polynomial in the input length.)

2. $(\varepsilon, 1 - \varepsilon)$-Fidelity-Pure-Pure is BQP-hard, even when $\varepsilon$ decays exponentially in the input length.

\textbf{Proof.} The containment of $(\alpha, \beta)$-Fidelity-Pure-Pure in BQP is a direct consequence of Algorithm 1.

So we focus on proving the hardness result. Consider an arbitrary problem $L$ in BQP. Thus, there exists a family $Q$ of circuits such that (146) and (147) hold. Given an instance $x$, the acceptance probability of the BQP algorithm is

$$p_{\text{acc}} = \|\langle 1|D \otimes I_G|Q|x\rangle_S\|_2^2$$

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\begin{align*}
\|\langle 1|D \otimes I_G|Q|x\rangle_S\|_2 &\geq \alpha(|x|). 
\end{align*}

(146)

2. Soundness: For all $x \in A_{\text{no}},$

$$\Pr[Q \text{ accepts } x] \leq b(|x|),$$

(147)

where, as clarified by the mathematical expression in (146), acceptance is defined as obtaining the outcome one upon measuring the decision qubit register $D$. BQP is then defined as BQP(2/3, 1/3).

1. Fidelity between two pure states

We now prove that the promise version of the problem of estimating the fidelity between two pure states is BQP-complete. In this problem and all that follows, the parameter $x$ is the description of the circuits involved, and the length $|x|$ is the number of bits needed to describe these circuits.

\textbf{Problem 1} ($(\alpha, \beta)$-Fidelity-Pure-Pure) Let $\alpha$ and $\beta$ be such that $0 \leq \alpha < \beta \leq 1$. Given are descriptions of circuits $U_S^\psi$ and $U_S^\phi$ that prepare the pure states $\psi_S$ and $\phi_S$, respectively. Decide which of the following holds.

\begin{align*}
\text{Yes:} &\quad F(\psi_S, \phi_S) \geq 1 - \alpha, \\
\text{No:} &\quad F(\psi_S, \phi_S) \leq 1 - \beta.
\end{align*}

(148)

(149)

Theorem 12 The promise problem Fidelity-Pure-Pure is BQP-complete.
Expanding $\text{CNOT}_{DC}$ as

$$\text{CNOT}_{DC} := |0\rangle\langle 0|_D \otimes I_C + |1\rangle\langle 1|_D \otimes X_C,$$

(152)

where $X_C$ denotes the Pauli-$X$ operator, it follows that

$$\tilde{p}_{\text{acc}} = \left| \langle x|S(0) A|1\rangle C \right|^2 \text{CNOT}_{DC} Q(\langle x|S(0) A|0\rangle C)^2.$$

(153)

Comparing this expression to (150), we see that the modified circuit has an acceptance probability equal to the square of the acceptance probability of the original BQP problem. Thus, by repeating the modified algorithm sufficiently many times, we can estimate the acceptance probability $\tilde{p}_{\text{acc}}$, and by taking a square root, we can output an estimate of the acceptance probability $p_{\text{acc}}$ of the original problem. In Appendix B, we derive the number of samples required to estimate $p_{\text{acc}}$ with accuracy $\varepsilon$ and error probability $\delta$.

The last step to be shown is that the modified acceptance probability $\tilde{p}_{\text{acc}}$ can be rewritten as the fidelity between two pure states. From (151), we see that

$$\tilde{p}_{\text{acc}} = \left| \langle x|S(0) A|1\rangle C \right|^2 \text{CNOT}_{DC} Q(\langle x|S(0) A|0\rangle C)^2.$$

(154)

where

$$|\psi\rangle := |x\rangle S(0) A|1\rangle C,$$

(155)

$$|\phi\rangle := Q^\dagger \text{CNOT}_{DC} Q|x\rangle S(0) A|0\rangle C.$$  

(156)

Thus, an arbitrary instance of a BQP problem can be rewritten as an instance of the fidelity between two pure states, proving that Fidelity-Pure-Pure is indeed a BQP-hard problem.

2. Fidelity between a pure state and a mixed state

Problem 2 ((α, β)-Fidelity-Pure-Mixed) Let $\alpha$ and $\beta$ be such that $0 < \alpha < \beta \leq 1$. Given are descriptions of circuits $U_{RS}^\alpha$ and $U_{RS}^\beta$ that prepare a purification of a mixed state $\rho_S$ and a pure state $\psi_S$, respectively. Decide which of the following holds.

Yes: $F(\rho_S, \psi_S) \geq 1 - \alpha$,  

(157)

No: $F(\rho_S, \psi_S) \leq 1 - \beta$.  

(158)

Theorem 13 The promise problem Fidelity-Pure-Mixed is BQP-complete.

1. $(\alpha, \beta)$-Fidelity-Pure-Mixed is in BQP for all $\alpha < \beta$. (It is implicit that the gap between $\alpha$ and $\beta$ is larger than an inverse polynomial in the input length.)

2. $(\varepsilon, 1 - \varepsilon)$-Fidelity-Pure-Mixed is BQP-hard, even when $\varepsilon$ decays exponentially in the input length.

Thus, $(\alpha, \beta)$-Fidelity-Pure-Mixed is BQP-complete for all $\alpha$ such that $0 < \alpha < \beta < 1$.

Proof. The containment of $(\alpha, \beta)$-Fidelity-Pure-Mixed in BQP is a direct consequence of Algorithm 3.

So we focus on proving the hardness result. Let $L$ be an arbitrary promise problem in BQP, and let $\{\phi_{DG}^x\}_x$ be a family of efficiently preparable pure states witnessing membership of $L$ in BQP. System $D$ is a decision qubit indicating acceptance or rejection of $x$, and system $G$ is a garbage system that purifies $D$. Suppose that the family $\{\phi_{DG}^x\}_x$ has completeness $1 - \delta$ and soundness $\delta$. If $x$ is a yes-instance of $L$, then, by the definition of BQP, it follows that $\|\langle 1|D|\phi_{DG}^x\rangle\|_2^2 \geq 1 - \delta$. On the other hand, if $x$ is a no-instance of $L$, then $\|\langle 1|D|\phi_{DG}^x\rangle\|_2^2 \leq \delta$. Since

$$\|\langle 1|D|\phi_{DG}^x\rangle\|_2 = F(|1|D, Tr_G[\phi_{DG}^x]|1_D)$$

(159)

it follows directly that this is an instance of $(1 - \delta, \delta)$-Fidelity-Pure-Mixed, given that the reduced state $Tr_G[\phi_{DG}^x]$ can be prepared efficiently, as well as the state $|1\rangle |1_D\rangle$. The desired hardness result then follows because BQP($c, s$) $\subseteq$ BQP($\delta, 1 - \delta$), for every $\delta$ exponentially small in the input length.

3. Hilbert–Schmidt distance

The next result we prove is that the promise version of the problem of estimating the normalized Hilbert–Schmidt distance of two arbitrary states is BQP-complete.

Recall that the normalized Hilbert–Schmidt distance of two states $\rho$ and $\sigma$ is given by

$$\frac{1}{\sqrt{2}} \|\rho - \sigma\|_2 := \frac{1}{\sqrt{2}} \sqrt{\text{Tr}[(\rho - \sigma)^2]}$$

(159)

$$= \frac{1}{\sqrt{2}} \sqrt{\text{Tr}[\rho^2] + \text{Tr}[\sigma^2] - 2\text{Tr}[\rho \sigma]}.$$  

(160)

If $\rho = \sigma$, then the Hilbert–Schmidt distance is equal to zero. The prefactor of $2^{-1/2}$ is the correct normalization by the following argument. Since $\text{Tr}[\rho \sigma] \geq 0$, the maximum value of the normalized distance satisfies

$$\frac{1}{\sqrt{2}} \sqrt{\text{Tr}[\rho^2] + \text{Tr}[\sigma^2] - 2\text{Tr}[\rho \sigma]}$$

$$\leq \frac{1}{\sqrt{2}} \sqrt{\text{Tr}[\rho^2] + \text{Tr}[\sigma^2]}$$

$$\leq 1,$$

(162)

where the second inequality follows because the purity of an arbitrary state $\rho$ satisfies $\text{Tr}[\rho^2] \leq 1$. The upper bound is achieved by pure orthogonal states.

Problem 3 ((\alpha, \beta)-Hilbert–Schmidt-Distance) Let $\alpha$ and $\beta$ be such that $0 < \alpha < \beta \leq 1$. Given are descriptions of circuits $U_{RS}^\alpha$ and $U_{RS}^\beta$ that prepare a purification of a mixed state $\rho_S$ and a pure state $\sigma_S$, respectively. Decide which of the following holds.

Yes: $\frac{1}{\sqrt{2}} \|\rho_S - \sigma_S\|_2 \geq 1 - \alpha,$

(163)
Theorem 14 The promise problem Hilbert–Schmidt-Distance is BQP-complete.

1. $(\alpha,\beta)$-Hilbert–Schmidt-Distance is in BQP for all $\alpha < \beta$. (It is implicit that the gap between $\alpha$ and $\beta$ is larger than an inverse polynomial in the input length.)

2. $(\varepsilon,1-\varepsilon)$-Hilbert–Schmidt-Distance is BQP-hard, even when $\varepsilon$ decays exponentially in the input length.

Thus, $(\alpha,\beta)$-Hilbert–Schmidt-Distance is BQP-complete for all $(\alpha,\beta)$ such that $0 < \alpha < \beta < 1$.

Proof. To show that the problem is BQP-complete, we need to demonstrate two facts: first, that the problem is in BQP, and second, that it is BQP-hard. Let us begin by proving that the problem is in BQP. This part of the proof is well known and understood by now, and it has been used in many quantum algorithms. We discuss it here for completeness. The intuitive idea is to estimate each term in (162) separately using a swap test. A term of the form $\text{Tr}[\rho \sigma]$, where $\rho$ and $\sigma$ are $n$-qubit states, can be estimated by repeatedly performing a swap test sufficiently many times to get a good estimate. Since there are only three terms to estimate, it follows that the problem is in BQP.

Next, we show that any problem in the BQP class can be reduced to this problem. A simpler way to show this is to map a known BQP-complete problem to our problem. We now show that the BQP-complete Fidelity-Pure-Pure problem can be reduced to this problem. A special case of the Hilbert–Schmidt-Distance problem is when both inputs are pure states. In this scenario, the normalized Hilbert–Schmidt distance is given by

$$\frac{1}{\sqrt{2}} \| \langle \psi | \psi \rangle - |\phi \rangle \langle \phi | \|_2 = \sqrt{1 - |\langle \psi | \phi \rangle|^2}.$$  (165)

Then the YES instance condition in (163) and (165) imply that $F(\psi, \phi) \leq \alpha(2-\alpha)$, in the case of a YES instance of Hilbert–Schmidt–Distance, and the NO instance condition in (164) and (165) imply that $F(\psi, \phi) \geq \beta(2-\beta)$, in the case of a NO instance of Hilbert–Schmidt–Distance. Since the function $x \rightarrow x(2-x)$ is a bijection on the unit interval $[0,1]$, it follows that the ability to decide Hilbert–Schmidt–Distance for pure states implies the ability to decide Fidelity-Pure-Pure, which is a BQP-complete problem by Theorem 12. We thus conclude that Hilbert–Schmidt–Distance is BQP-Hard. This, along with the fact that the problem is in the BQP class, concludes the proof.

Remark 15 The normalized Schatten-$p$ distance between two states $\rho$ and $\sigma$ is defined as

$$\frac{1}{2^{1/p}} \| \rho - \sigma \|_p := \frac{1}{2^{1/p}} (\text{Tr}[|\rho - \sigma|^p])^{1/p}. $$  (166)

We can formulate promise problems from these quantities, generalizing Hilbert–Schmidt–Distance in Problem 3. Plugging pure states $\psi$ and $\phi$ into (166) and exploiting the fact that the eigenvalues of $\psi - \phi$ are equal to $|\sin \theta|$ and $-|\sin \theta|$ [Wil17, Proof of Theorem 9.3.1], where $\theta$ satisfies $F(\psi, \phi) = \cos^2 \theta$, it follows that

$$\frac{1}{2^{1/p}} \| \psi - \phi \|_p = \sqrt{1 - F(\psi, \phi)}$$  (167)

for all $p \geq 1$. Thus, by the same reasoning given in the second part of the proof of Theorem 14, we conclude that these promise problems are all BQP-hard.

Now consider that estimating the Schatten-2k distance between two states, where $k \in \mathbb{N}$, is in BQP. For constant $k$, each term in the expansion of $\| \rho - \sigma \|_{2k}^k = \text{Tr}[|\rho - \sigma|^{2k}]$ can be estimated in polynomial time [EAO+02], in fact in constant quantum depth [QKW22] after the circuits that prepare multiples copies of $\rho$ and $\sigma$ are executed. Thus, combining with the above, we conclude that, for each constant $k \in \mathbb{N}$, the promise version of the problem of estimating $\frac{1}{2^{1/p}} \| \rho - \sigma \|_{2k}$ is a BQP-complete problem.

### B. Fidelity between a pure state and a channel (QMA-complete)

Next, we provide a proof that the promise version of the problem of evaluating the fidelity between a channel and a pure state is QMA-complete. The definition of QMA can be found in [Wat09a], reproduced here for convenience (but again slightly different in that we consider unitary circuits). Let $A = (A_{\text{yes}}, A_{\text{no}})$ be a promise problem, let $p, q$ be polynomially-bounded functions, and let $a, b : \mathbb{N} \rightarrow [0,1]$ be functions. Then $A \in \text{QMA}_p(a,b)$ if there exists a polynomial-time generated family of unitary circuits $Q = \{Q_n : n \in \mathbb{N}\}$, where each circuit $Q_n$ takes $n + p(n) + q(n)$ input qubits and produces one decision qubit $D$ and $n + p(n) + q(n) - 1$ garbage qubits $G$, with the following properties (as before, we abbreviate each $Q_n$ as $Q_{\text{sap} \rightarrow DG}$, thereby suppressing the dependence on the input length $n = |x|$ and explicitly indicating the systems involved at the input and output of the unitary):

1. Completeness: For all $x \in A_{\text{yes}}$, there exists a $q(|x|)$-qubit quantum state $\sigma$ such that

$$\Pr[Q \text{ accepts } (x, \sigma)] = (1\langle D \mid \text{Tr}_G[\omega_{DG}]1\rangle_D \geq a(|x|).$$  (169)

where $\omega_{DG} := Q_{\text{sap} \rightarrow DG}(|x\rangle\langle x|_S \otimes |0\rangle_0 |\text{a} \otimes \sigma_P)(Q_{\text{sap} \rightarrow DG})$.  (170)

2. Soundness: For all $x \in A_{\text{no}}$, and every $q(|x|)$-qubit quantum state $\sigma$, the following inequality holds:

$$\Pr[Q \text{ accepts } (x, \sigma)] \leq b(|x|).$$  (171)


Then QMA = \( \bigcup_p \text{QMA}_p \{(2/3,1/3) \} \), where the union is over every polynomially bounded function \( p \).

**Problem 4 ((\( \alpha, \beta \))-Fidelity-Channel-Pure)** Let \( \alpha \) and \( \beta \) be such that \( 0 \leq \alpha < \beta \leq 1 \). Given are descriptions of circuits \( U_{SR \to BE}^N \) and \( U_B^N \) that prepare a unitary dilation of a channel

\[
N_{S \to B}(\cdot) := \text{Tr}_E[U_{SR \to BE}^N(\cdot)|S \otimes |0\rangle|0\rangle_R(U_{SR \to BE}^N)^\dagger]
\]

and a pure state \( \psi_B := U_B^N|0\rangle|0\rangle_B(U_B^N)^\dagger \), respectively. Decide which of the following holds:

- **Yes:** \( \max_{\rho_S} F(N_{S \to B}(\rho_S), \psi_B) \geq 1 - \alpha \), (173)
- **No:** \( \max_{\rho_S} F(N_{S \to B}(\rho_S), \psi_B) \leq 1 - \beta \), (174)

where the maximization is over every input density operator \( \rho_S \).

**Theorem 16** The promise problem Fidelity-Channel-Pure is QMA-complete.

1. \((\alpha, \beta)\)-Fidelity-Channel-Pure is in QMA for all \( \alpha < \beta \). (It is implicit that the gap between \( \alpha \) and \( \beta \) is larger than an inverse polynomial in the input length.)
2. \((\varepsilon, 1 - \varepsilon)\)-Fidelity-Channel-Pure is QMA-hard, even when \( \varepsilon \) decays exponentially in the input length.

Thus, \((\alpha, \beta)\)-Fidelity-Channel-Pure is QMA-complete for all \((\alpha, \beta)\) such that \( 0 < \alpha < \beta < 1 \).

**Proof.** To show that the problem is QMA-complete, we need to demonstrate two facts: first, that the problem is in QMA, and second, that it is QMA-hard.

Let us begin by proving that the problem is in QMA. The intuitive idea is that the prover sends an optimal state \( \rho_S \) to the verifier, who then performs the channel \( N_{S \to B} \) on it, followed by the unitary \((U_B^N)^\dagger\). The verifier then performs a computational basis measurement on all registers of system \( B \) and accepts if and only if the all-zeros measurement outcome occurs. Indeed, the acceptance probability of this scheme is precisely equal to the fidelity in (173):

\[
\langle \psi | N_{S \to B}(\rho_S) | \psi \rangle = F(N_{S \to B}(\rho_S), \psi_B). \tag{175}
\]

To bring the original expression more closely to the form given in (168), observe that

\[
\begin{align*}
&\langle 0_b | U_B^N | N_{S \to B}(\rho_S) | U_B^N | 0_B \rangle = \langle 0_b | X_B(U_B^N)^\dagger \times \\
&\text{Tr}_E[U_{SR \to BE}^N(0_b \otimes \rho_S)(U_{SR \to BE}^N)^\dagger | U_B^N | X_B | 1_b \rangle, \tag{176}
\end{align*}
\]

where \( X_B \) is understood to be the tensor power Pauli \( X \) operator acting on all qubits of the \( B \) register. To bring the final expression exactly into the form in (168), we need a single decision qubit that we measure. We can use a multi-controlled Toffoli gate from the \( B \) register to a single qubit decision qubit. Thus, if we identify \( x \) with \( 0 \), \( x \) with \( \rho_S \), and \( q \) with \( (X_B \otimes I_E) \circ ((U_B^N)^\dagger \otimes I_E) \circ U_{SR \to BE}^N \), it follows that the problem belongs to the QMA class.

Next, we show that any problem in the QMA class can be polynomially reduced to this problem. Let \( P \) be an arbitrary problem in the QMA class. This implies that (168) and (171) must hold. This problem can then be thought of as a fidelity problem with a channel \( M_x \) defined as

\[
M_{SAP \to D}(\cdot) := \text{Tr}_G[|x\rangle\langle x|_S \otimes |0\rangle\langle 0|_A \otimes (\cdot)]Q^1]. \tag{177}
\]

Furthermore, we identify the state \( \psi \) from the fidelity problem with \( |1\rangle|1\rangle_D \), and then we find that

\[
\langle 1_D | \text{Tr}_G[|x\rangle\langle x|_S \otimes |0\rangle\langle 0|_A \otimes (\cdot)]Q^1] | 1_D \rangle = F(M^x, |1\rangle|1\rangle). \tag{179}
\]

It follows directly that this is an instance of \((1 - a(|x|), 1 - b(|x|))\)-Fidelity-Channel-Pure, given that the channel \( M_x \) can be prepared efficiently, as well as the state \(|1\rangle|1\rangle \). The desired hardness result then follows because \( \text{QMA}(1 - a(|x|), 1 - b(|x|)) \subseteq \text{QMA}(\delta, 1 - \delta) \), for every \( \delta \) exponentially small in the input length. \( \blacksquare \)

**C. Fidelity between a pure state and a channel with separable input (QMA(2)-complete)**

Lastly, we provide a proof for the result that the promise version of the problem of evaluating the fidelity between a pure state and a channel with a separable state as input is QMA(2)-complete. A state is separable if and only if it is not entangled. A separable state \( \sigma_{SR} \) can be expanded as follows:

\[
\sigma_{SR} = \sum_k p(k) |\phi^k\rangle \langle \phi^k|_S \otimes |\phi^k\rangle \langle \phi^k|_R, \tag{180}
\]

where \( \{p(k)\}_k \) is a probability distribution and \( \{|\phi^k\rangle \langle \phi^k|_S\}_k \) and \( \{|\phi^k\rangle \langle \phi^k|_R\}_k \) are sets of pure states. SEP is defined as the set of all separable states. QMA(2) is a generalization of QMA with proofs that consist of two systems guaranteed to be unentangled [KMY01, HM10].

We reproduce the definition of QMA(2) for convenience.

Let \( A = (A_{sys}, A_{con}) \) be a promise problem, let \( p, q, r \) be polynomially-bounded functions, and let \( a, b : \mathbb{N} \to [0, 1] \) be functions. Then \( A \in \text{QMA}(2)_{p,q}(a, b) \) if there exists a polynomial-time generated family of circuits \( Q = \{Q_n : n \in \mathbb{N}\} \), where each circuit \( Q_n \) takes \( n + p(n) + q(n) + r(n) \) input qubits and produces one decision qubit \( D \) and \( n + p(n) + q(n) + r(n) - 1 \) garbage
qubits $G$, with the following properties (again, we employ the notation $Q_{SAP, P, D}$ in what follows):

1. Completeness: For all $x \in A_{\text{yes}}$, there exists a $q(|x|)$-qubit state $\rho$ and an $r(|x|)$-qubit state $\sigma$ such that
   \[
   \Pr[Q \text{ accepts } (x, \rho, \sigma)] = \langle 1 | \rho \mathrm{Tr}_G[\omega_{DG}] | 1 \rangle_D \geq a(|x|),
   \]  

   where
   \[
   \omega_{DG} := Q_{SAP, P, D}([x|x]_S \otimes |0\rangle_0 A \otimes \rho_p \otimes \sigma_p)(Q_{SAP, P, D}).
   \]  

2. Soundness: For all $x \in A_{\text{no}}$, and every $q(|x|)$-qubit state $\rho$ and $r(|x|)$-qubit state $\sigma$, the following inequality holds:
   \[
   \Pr[Q \text{ accepts } (x, \rho, \sigma)] \leq b(|x|).
   \]

Then $QMA(2) = \bigcup_q QMA(2)_{p,q} (2/3, 1/3)$, where the union is over all polynomially bounded functions $p$ and $q$.

**Problem 5 ((\(\alpha, \beta\))-Fidelity-Pure-Channel-Sep-Inp)**

Let $\alpha$ and $\beta$ be such that $0 < \alpha < \beta < 1$. Given are descriptions of circuits $U^{N}_{SRE \rightarrow AE', A}$ and $U^{N}_{A}$ that prepare a unitary dilation of a channel

\[
N_{SR \rightarrow A} := \mathrm{Tr}_E[U^{N}_{SRE \rightarrow AE'}(|\cdot|_S \otimes |0\rangle_0 E)(U^{N}_{SRE \rightarrow AE'})^\dagger],
\]

and a pure state $\psi_A$, respectively. Decide which of the following holds:

**Yes:** \[\max_{\sigma_{SR} \in \text{SEP}} F(N_{SR \rightarrow A}(\sigma_{SR}), \psi_A) \geq 1 - \alpha, \]  

**No:** \[\max_{\sigma_{SR} \in \text{SEP}} F(N_{SR \rightarrow A}(\sigma_{SR}), \psi_A) \leq 1 - \beta. \]

**Theorem 17** The promise problem Fidelity-Pure-Channel-Sep-Inp is $QMA(2)$-complete.

1. $(\alpha, \beta)$-Fidelity-Pure-Channel-Sep-Inp is in $QMA(2)$ for all $\alpha < \beta$. (It is implicit that the gap between $\alpha$ and $\beta$ is larger than an inverse polynomial in the input length.)

2. $(\varepsilon, 1 - \varepsilon)$-Fidelity-Pure-Channel-Sep-Inp is $QMA(2)$-hard, even when $\varepsilon$ decays exponentially in the input length.

Thus, $(\alpha, \beta)$-Fidelity-Pure-Channel-Sep-Inp is $QMA(2)$-complete for all $(\alpha, \beta)$ such that $0 < \alpha < \beta < 1$.

**Proof.** To show that the problem is $QMA(2)$-complete, we need to demonstrate two facts: first, that the problem is in $QMA(2)$, and second, that it is $QMA(2)$-hard. Let us begin by proving that the problem is in $QMA(2)$. The intuitive idea is that the two provers, using shared randomness, send an optimal separable state $\sigma_{SR}$ to the verifier, who then performs the channel $N_{SR \rightarrow A}$ on it, followed by the unitary $(U^{N}_{A})^\dagger$. (Note that $QMA(2)$ remains unchanged if the provers have access to shared randomness [HM10].) The verifier then performs a computational basis measurement on all registers of system $A$ and accepts if and only if the all-zeros measurement outcome occurs.

Consider that a separable state can be decomposed as

\[
\sigma_{SR} = \sum_{k} p(k) |\phi^k_S\rangle \langle \phi^k_S| \otimes |\phi^k_R\rangle \langle \phi^k_R|.
\]

Indeed, the acceptance probability of this scheme is precisely equal to the fidelity in (185):

\[
F(N_{SR \rightarrow A}(\sigma_{SR}), \psi_A) = \langle \psi \rangle_A N_{SR \rightarrow A}(\sigma_{SR})| \psi_A = \sum_{k} p(k) \langle \psi \rangle_A N_{SR \rightarrow A}(|\alpha^k_S\rangle \langle \alpha^k_S| \otimes |\phi^k_R\rangle \langle \phi^k_R|)| \psi_A.
\]

The final expression is an average of individual elements. Thus, taking a maximization over all separable states and noting that the maximum is always greater than the average, we conclude that

\[
\max_{\sigma_{SR} \in \text{SEP}} F(N_{SR \rightarrow A}(\sigma_{SR}), \psi_A) = \max_{|\phi_S, \phi_R\rangle} \langle \psi \rangle_A N_{SR \rightarrow A}| \psi_A.
\]

Thus, we see that

\[
\max_{\sigma_{SR} \in \text{SEP}} F(N_{SR \rightarrow A}(\sigma_{SR}), \psi_A) = \max_{|\phi_S, \phi_R\rangle} \langle \psi \rangle_A N_{SR \rightarrow A} \times (U^{N}_{A})^\dagger \mathrm{Tr}_E[U^{N}_{SRE \rightarrow AE'}(|0\rangle_0 E \otimes \phi_S \otimes \phi_R) \times (U^{N}_{SRE \rightarrow AE'})^\dagger |\psi_A\rangle_A.
\]

where $X_A$ is understood to be the tensor-power Pauli $X$ operator acting on all qubits of the $A$ register. To bring the final expression into the precise form in (181), we need a single decision qubit that we measure. We can use a multi-controlled Toffoli gate from the $A$ register to a single qubit decision qubit. Thus, if we identify $x$ with $0$, $\rho$ with $\varphi_S$, $\sigma$ with $\phi_R$ and $Q_n$, with $(X_A \otimes I_R) \circ (U^{N}_{A})^\dagger \otimes I_R) \circ U^{N}_{SRE \rightarrow AE'}$, it follows that the problem belongs to the $QMA(2)$ class.

Next, we show that any problem in the $QMA(2)$ class can be polynomially reduced to this problem. Let $P$ be an arbitrary problem in the $QMA(2)$ class. This implies that (181) and (183) must hold. This problem can then be thought of as a fidelity problem with a channel $\mathcal{M}_x$ defined as

\[
\mathcal{M}_{SAP, P, D}^{x} := \mathrm{Tr}_G[Q_n(|x|_S \otimes |0\rangle_0 A \otimes \cdot |p_1, p_2) Q_n^\dagger].
\]
Furthermore, by identifying the state $\psi$ from the fidelity problem with $|1\rangle|1\rangle$, then we find that
\begin{align}
(1|\text{Tr}_G [Q(|x\rangle\langle x|)S \otimes |0\rangle\langle 0|_A \otimes \psi_1 \otimes \psi_2]Q^\dagger]|1\rangle) \\
= (1|M_\delta(\psi_1 \otimes \psi_2)|1\rangle \\
= F(M_\delta(\psi_1 \otimes \psi_2), |1\rangle|1\rangle).
\end{align}

It follows directly that this is an instance of $(1 - a(|x\rangle), 1 - b(|x\rangle))$-Fidelity-Channel-Pure, given that the channel $M_\delta$ can be prepared efficiently, as well as the state $|1\rangle|1\rangle$. The desired hardness result then follows because QMA$(1 - a(|x\rangle), 1 - b(|x\rangle)) \subseteq \text{QMA}(\delta, 1 - \delta)$, for every $\delta$ exponentially small in the input length (see [HM10, Theorem 9]).

\section{VI. Generating Fixed Points of Quantum Channels}

In this section, we discuss how Algorithm 10 can generate a fixed-point state or an approximate fixed-point state of a quantum channel. There are various associated subtleties in such a scenario that we consider.

As a special case of Algorithm 10, we can select $N^0_{A \rightarrow B}$ to be a channel $N$ with its output and input systems having the same dimension (i.e., $|A| = |B|$), and we can select the second channel $N^0_{A \rightarrow B}$ to be the identity channel. In this case, the quantity in (78) is always equal to one. This follows from the well known fact that every quantum channel with matching input and output systems has a fixed point state [EHK78] (see also [Deu91, Wol12]) and because the prover’s goal is to maximize the acceptance probability. That is, for every such channel $N$, there exists a state $\rho$ such that
\begin{equation}
N(\rho) = \rho,
\end{equation}
and so the prover can simply send this state. Related to this, there is a faithfulness property that holds. If the acceptance probability is equal to one, then it follows that
\begin{equation}
\sup_{\rho} F(N(\rho), \rho) = 1,
\end{equation}
and we conclude that there exists a state $\rho$ satisfying (194) because the fidelity is continuous and the set of density operators is convex and compact.

What is interesting in this case is that Algorithm 10 outputs a fixed point of the channel $N$. Fixed points of quantum channels are important not only for understanding thermalization in a physical process [BCL+21] (a fixed point can be understood as an equilibrium state of the channel) but also in the Deutschian theory of closed timelike curves [Deu91].

We can also modify this approach slightly and employ Algorithm 13. In this case, the verifier can employ the following ensemble of channels
\begin{equation}
\left\{ \frac{1}{L} N^\ell \right\}_{\ell = 0}^{L - 1},
\end{equation}
where $N^\ell$ here is defined as
\begin{equation}
N^\ell = N \circ \cdots \circ N,
\end{equation}
in this case, the acceptance probability of Algorithm 13 is given by
\begin{equation}
\frac{1}{L} \sup_{\rho, \sigma} \sum_{\ell = 0}^{L - 1} \sqrt{F(N^\ell(\rho), \sigma)}^2.
\end{equation}
This is again equal to one because the prover can transmit a fixed point to the verifier, which satisfies
\begin{equation}
N^\ell(\rho) = \rho \quad \forall \ell \in \{0, \ldots, L - 1\}.
\end{equation}

Similarly, in this case, a faithfulness property holds as well. If the expression in (198) is equal to one, then there exists a state $\rho$ satisfying (199). Furthermore, Algorithm 13 outputs a fixed point satisfying (198).

The cases outlined above are simple. The situation becomes more subtle when the verifier tries to use the state sent by the prover to solve a computational problem, as is the case in quantum computation in the presence of Deutschian closed timelike curves [AW09]. In this case, there are different goals, which are 1) to pass the test of the verifier in Algorithm 13, as well as 2) to have the decision qubit be as close as possible to the $|1\rangle|1\rangle$ state. In this case, the prover need not send an exact fixed point, but only send an approximate fixed point, satisfying
\begin{equation}
F(\rho, N(\rho)) \geq 1 - \varepsilon,
\end{equation}
or
\begin{equation}
\frac{1}{L} \sup_{\sigma} \sum_{\ell = 0}^{L - 1} \sqrt{F(N^\ell(\rho), \sigma)}^2 \geq 1 - \varepsilon,
\end{equation}
where $\varepsilon \in (0, 1)$. The prover can do this to optimize the overall acceptance probability of the QIP algorithm. Somewhat counter-intuitively, approximate fixed points need not be close to exact fixed points, as illustrated by the following example. Suppose that $N$ is a classical channel that takes $1 \rightarrow 1$ deterministically, but then takes $0 \rightarrow 0$ with probability $1 - \varepsilon$ and $0 \rightarrow 1$ with probability $\varepsilon$. In this case, $1$ is the exact fixed point of this stochastic process, but $0$ is an approximate fixed point satisfying (200). However, $0$ is completely distinguishable from $1$ (the fidelity of these two classical states is equal to zero).

In Appendix C, we discuss various issues related to fixed points and approximate fixed points of quantum channels when attempting to understand quantum interactive proofs and the computational complexity of Deutschian closed timelike curves.

\section{VII. Conclusion}

In this paper, we have delineated several algorithms for estimating distinguishability measures on quantum computers. All of the measures are based on trace distance or
fidelity, and we have considered them for quantum states, channels, and strategies. Many of the algorithms rely on interaction with a quantum prover, and in these cases, we have replaced the prover with a parameterized quantum circuit. As such, these methods are not guaranteed to converge for all possible states, channels, and strategies. It is an interesting open question to determine conditions under which the algorithms are guaranteed to converge and run efficiently.

We have also simulated several of the algorithms in both the noiseless and noisy scenarios. We found that the simulations converge well for all states and channels considered, and for all algorithms simulated. As more advanced quantum computers become available (with more qubits and greater reliability), it would be interesting to simulate our algorithms for states and channels involving larger numbers of qubits. All of our Python code is written in a modular way, such that it will be straightforward to explore this direction. Lastly, we proved several complexity-theoretic results about various distance estimation algorithms; in particular, we showed and, in some cases, recalled that there is a fidelity or distance estimation problem that is complete for the commonly studied complexity classes BQP, QMA, QMA(2), QSZK, QIP(2), and QIP.

Going forward from here, it remains open to determine methods for estimating other distinguishability measures such as the Petz–Rényi relative entropy [Pet85, Pet86] and the sandwiched Rényi relative entropy [MLDS+13, WWY14] of channels [LKDW18] and strategies [WW19]. More generally, one could consider distinguishability measures beyond these. One desirable aspect of the algorithms appearing in this paper is that they provide a one-shot interpretation for the various distinguishability measures as the maximum acceptance probability in a quantum interactive proof (with the trace-distance based algorithms and interpretations being already known from [Wat02b, RW05, GW07, Gut09, Gut12]). However, it is unclear to us whether one could construct a quantum interactive proof for which the maximum acceptance probability is related to the Petz– or sandwiched Rényi relative entropy of a channel or a strategy.

Note added: While finalizing the results of our initial arXiv post [ARSW21], we noticed the arXiv post [BBC21], which is related to the contents of Section III. Ref. [BBC21] is now published as [BBC22].

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Appendix A: Proofs from main text

1. Proof of Theorem 1

**Proof of Theorem 1.** After Step 1 of Algorithm 4, the global state is

$$|\Phi_T\rangle_T = |0\rangle_{RS}. \quad (A1)$$
After Step 2 of Algorithm 4, it is
\[
\frac{1}{\sqrt{2}} \left\lVert \sum_{i \in \{0,1\}} \langle i | i \rangle T | i \rangle | \psi^i \rangle_R S \right\rVert_f .
\] (A2)

After Step 4 of Algorithm 4, it is
\[
P_{T^*R^F \to T^*F^*} \left( \frac{1}{\sqrt{2}} \sum_{i \in \{0,1\}} \langle i | i \rangle T | i \rangle | \psi^i \rangle_R S | 0 \rangle_F \right) .
\] (A3)

For a fixed unitary \( P = P_{T^*R^F \to T^*F^*} \) of the prover, the acceptance probability is then
\[
\left\lVert \langle \Phi | T^*TP \left( \frac{1}{\sqrt{2}} \sum_{i \in \{0,1\}} \langle i | i \rangle T | i \rangle | \psi^i \rangle_R S | 0 \rangle_F \right) \right\rVert^2_f
= \frac{1}{2} \left\lVert \langle \Phi | T^*TP \sum_{i \in \{0,1\}} \langle i | T | \psi^i \rangle_R S | 0 \rangle_F \right\rVert^2_f .
\] (A4)

In a quantum interactive proof, the prover is trying to maximize the probability that the verifier accepts. So the acceptance probability of Algorithm 4 is given by
\[
\max_{P_{T^*R^F \to T^*F^*}} \frac{1}{2} \left\lVert \langle \Phi | T^*TP \sum_{i \in \{0,1\}} \langle i | T | \psi^i \rangle_R S | 0 \rangle_F \right\rVert^2_f .
\] (A5)

Setting
\[
P_{R^0 \to F^*}^0 := | 0 \rangle_T^* P_{T^*R^F \to T^*F^*} | 0 \rangle_T | 0 \rangle_F ,
\] (A6)
\[
P_{R^1 \to F^*}^1 := | 1 \rangle_T^* P_{T^*R^F \to T^*F^*} | 1 \rangle_T | 0 \rangle_F ,
\] (A7)
we have that
\[
\frac{1}{2} \left\lVert \langle \Phi | T^*TP \sum_{i \in \{0,1\}} \langle i | T | \psi^i \rangle_R S | 0 \rangle_F \right\rVert^2_f
= \frac{1}{4} \left\lVert \sum_{i \in \{0,1\}} P_{R^i \to F^*}^i | \psi^i \rangle_R S \right\rVert^2_f
= \frac{1}{4} \sum_{i,j \in \{0,1\}} \langle \psi^j | R S (P_{R^i \to F^*}^i) | P_{R^j \to F^*}^j | \psi^i \rangle_R S
\leq \frac{1}{2} \left( 1 + \text{Re} \left\{ | \langle \psi^0 | R S (P_{R^0 \to F^*}^0) | P_{R^1 \to F^*}^1 | \psi^1 \rangle_R S \right\} \right)
\leq \frac{1}{2} \left( 1 + | \langle \psi^0 | R S (P_{R^0 \to F^*}^0) | P_{R^1 \to F^*}^1 | \psi^1 \rangle_R S \right) .
\] (A10)

The first inequality follows because \( P_{R^i \to F^*}^i \) is a contraction for \( i \in \{0,1\} \), so that \( (P_{R^i \to F^*}^i) | P_{R^i \to F^*}^i \leq I_{F^*} \). Then consider that
\[
| \langle \psi^0 | R S (P_{R^0 \to F^*}^0) | P_{R^1 \to F^*}^1 | \psi^1 \rangle_R S |
\leq \max_{P^0, P^1} \left\{ | \langle \psi^0 | R S (P_{R^0 \to F^*}^0) | P_{R^1 \to F^*}^1 | \psi^1 \rangle_R S : \left\lVert P^i \right\rVert_\infty \leq 1 \forall i \right\} .
\] (A12)

The last line is a consequence of the following reasoning (which is the same as that employed in Section III in [CHM+16]). The inequality
\[
\max_{P^0, P^1} \left\{ | \langle \psi^0 | R S (P_{R^0 \to F^*}^0) | P_{R^1 \to F^*}^1 | \psi^1 \rangle_R S | \right\} : \left\lVert P^i \right\rVert_\infty \leq 1 \forall i
\geq \sqrt{F(\rho^0_S, \rho^1_S)} .
\] (A14)
holds because the isometries \( P_{R^0 \to F^*}^0 \) and \( P_{R^1 \to F^*}^1 \) that achieve the maximum for the fidelity are each contractions and the optimization is conducted over all contractions. The opposite inequality
\[
\max_{P^0, P^1} \left\{ | \langle \psi^0 | R S (P_{R^0 \to F^*}^0) | P_{R^1 \to F^*}^1 | \psi^1 \rangle_R S | \right\} : \left\lVert P^i \right\rVert_\infty \leq 1 \forall i
\leq \sqrt{F(\rho^0_S, \rho^1_S)} .
\] (A15)
is a consequence of the fact that every contraction can be written as a convex combination of isometries [Zha11, Theorem 5.10]. Indeed, this means that, for each \( i \in \{0,1\} \),
\[
P_{R^i \to F^*} = \sum_x p_i(x) W_{R^i \to F^*}^i ,
\] (A16)
where \( \{ p_i(x) \}_x \) is a probability distribution and \( W_{R^i \to F^*}^i \) is an isometry, for each \( i \) and \( x \). Then we find that
\[
| \langle \psi^0 | R S (P_{R^0 \to F^*}^0) | P_{R^1 \to F^*}^1 | \psi^1 \rangle_R S |
= \left| \langle \psi^0 | R S \left( \sum_x p_0(x) W_{R^0 \to F^*}^0 \right)^* \times \left( \sum_x p_1(x') W_{R^1 \to F^*}^1 \right) | \psi^1 \rangle_R S \right|
\leq \sum_{x,x'} p_0(x) p_1(x') \left| \langle \psi^0 | R S \left( W_{R^0 \to F^*}^0 \right)^* W_{R^1 \to F^*}^1 | \psi^1 \rangle_R S \right|
\leq \sum_{x,x'} \left| \langle \psi^0 | R S \left( W_{R^0 \to F^*}^0 \right)^* W_{R^1 \to F^*}^1 | \psi^1 \rangle_R S \right|
\leq \sqrt{F(\rho^0_S, \rho^1_S)} .
\] (A21)

Thus, an upper bound on the acceptance probability of Algorithm 4 is as follows:
\[
\frac{1}{2} \left( 1 + \sqrt{F(\rho^0_S, \rho^1_S)} \right) .
\] (A22)

This upper bound can be achieved if the prover applies a unitary extension of the following isometry:
\[
P_{T^*R^F \to T^*F^*} = \sum_{i \in \{0,1\}} | i \rangle_T^* \otimes P_{R^i \to F^*} \otimes | 0 \rangle_F ,
\] (A23)

where \( P_{R^0 \to F^*} \) and \( P_{R^1 \to F^*} \) are isometries achieving the maximum in the fidelity \( F(\rho^0_S, \rho^1_S) \) .
2. Proof of Theorem 2

Proof of Theorem 2. After Step 1 of Algorithm 5, the global state is
\[ |\Phi\rangle_{TT} |0\rangle_{R_1S_1R_2S_2}. \tag{A24} \]

After Step 2, the global state is
\[ |\Phi\rangle_{TT'} |\psi^{(0)}\rangle_{R_1S_1} |\psi^{(1)}\rangle_{R_2S_2}. \tag{A25} \]

After Step 3, it becomes
\[ \frac{1}{\sqrt{2}} (|0\rangle_T |0\rangle_T |\psi^{(0)}\rangle_{R_1S_1} |\psi^{(1)}\rangle_{R_2S_2}) \]
\[ + \frac{1}{\sqrt{2}} (|1\rangle_T |1\rangle_T |\psi^{(0)}\rangle_{R_1S_1} |\psi^{(1)}\rangle_{R_1S_2}). \tag{A26} \]

The verifier then sends systems \( T' \), \( R_1 \), and \( R_2 \) to the prover, who appends the state \( |0\rangle_F \) and acts with a unitary \( P_{T'R_1R_2F'\rightarrow T'\psi^{(0)}} \). Without loss of generality, and for simplicity of the ensuing analysis, we can imagine that before applying the unitary \( P_{T'R_1R_2F'\rightarrow T'\psi^{(0)}} \), the prover applies a controlled SWAP to systems \( T' \), \( R_1 \), and \( R_2 \), so that the state before applying \( P_{T'R_1R_2F'\rightarrow T'\psi^{(0)}} \) is as follows:
\[ \frac{1}{\sqrt{2}} (|0\rangle_T |0\rangle_T |\psi^{(0)}\rangle_{R_1S_1} |\psi^{(1)}\rangle_{R_2S_2}) \]
\[ + \frac{1}{\sqrt{2}} (|1\rangle_T |1\rangle_T |\psi^{(1)}\rangle_{R_1S_1} |\psi^{(0)}\rangle_{R_2S_2}). \tag{A27} \]

This follows because the prover can apply arbitrary unitaries to his received systems, and one such possible unitary is to apply this controlled SWAP, undo it, and then apply \( P_{T'R_1R_2F'\rightarrow T'\psi^{(0)}} \). However, the latter two unitaries are a particular example of a unitary \( P_{T'R_1R_2F'\rightarrow T'\psi^{(0)}} \), so we proceed with the ensuing analysis assuming that the global state, before the prover applies \( P_{T'R_1R_2F'\rightarrow T'\psi^{(0)}} \), is given by \( (A27) \). Note that the actions of tensoring in the state \( |0\rangle_F \) and applying \( P_{T'R_1R_2F'\rightarrow T'\psi^{(0)}} \) together constitute an isometry
\[ P_{T'R_1R_2F'\rightarrow T'\psi^{(0)}} := P_{T'R_1R_2F'\rightarrow T'\psi^{(0)}} |0\rangle_F, \tag{A28} \]
resulting in the state
\[ \frac{1}{\sqrt{2}} P_{T'R_1R_2F'\rightarrow T'\psi^{(0)}} |0\rangle_T |0\rangle_T |\psi^{(0)}\rangle_{R_1S_1} |\psi^{(0)}\rangle_{R_2S_2} \]
\[ + \frac{1}{\sqrt{2}} P_{T'R_1R_2F'\rightarrow T'\psi^{(0)}} |1\rangle_T |1\rangle_T |\psi^{(0)}\rangle_{R_1S_1} |\psi^{(0)}\rangle_{R_2S_2}. \tag{A29} \]

Let us set
\[ P_{R_1R_2F'\rightarrow F'} := |0\rangle_T P_{T'R_1R_2F'\rightarrow T'\psi^{(0)}} |0\rangle_T, \tag{A30} \]
\[ P_{R_1R_2F'\rightarrow F'} := |1\rangle_T P_{T'R_1R_2F'\rightarrow T'\psi^{(0)}} |1\rangle_T. \tag{A31} \]

The verifier finally performs a Bell measurement and accepts if and only if the outcome \( \Phi_{T'\psi^{(0)}} \) occurs. The acceptance probability is then
\[ \frac{1}{2} \left( 1 + \sqrt{F(\rho^{(0)} \otimes \rho^{(1)} \otimes \rho^{(0)})} \right). \tag{A39} \]

The steps given above follow for reasons very similar to those given in the proof of Theorem 1. Continuing, we find that
\[ \text{Eq. (A38)} = \frac{1}{2} \left( 1 + \sqrt{F(\rho^{(0)} \otimes \rho^{(1)} \otimes \rho^{(0)})} \right). \tag{A39} \]
where we used the multiplicativity of the fidelity for tensor-product states to get (A40) and the symmetric property of fidelity to arrive at (A41). Thus, we have established (25) as an upper bound on the acceptance probability. This upper bound can be achieved by setting $F' \simeq R_1R_2$ and

$$P_{T',R_1R_2F \rightarrow T''E'} = |0\rangle_{T'} \langle 0|_{T'} \otimes I_{R_1R_2F} \otimes |0\rangle_F \quad (A42)$$
$$+ |1\rangle_{T'} \langle 1|_{T'} \otimes U_{R_1} \otimes U_{R_2} \otimes |0\rangle_F, \quad (A43)$$

where $U_{R_i}$ is a unitary that achieves the fidelity for $F(\rho^0, \rho^1)$, so that

$$\sqrt{F(\rho^0, \rho^1)} = \langle \psi^0 | R_1S_1U_{R_1} | \psi^0 \rangle_{R_1S_1}. \quad (A44)$$

This concludes the proof. ■

3. Proof of Theorem 3

Proof of Theorem 3. After Step 1 of Algorithm 8, the global state is

$$|\Phi\rangle_{T'T'} |\psi\rangle_{RA} |0\rangle_{E'}. \quad (A45)$$

After Step 2 of Algorithm 8, it is

$$\frac{1}{\sqrt{2}} \sum_{i \in \{0,1\}} |i\rangle_T |i\rangle_T U^i |\psi\rangle_{RA} |0\rangle_{E'}, \quad (A46)$$

where $U^i = U^i_{AE'} \rightarrow BE$ for $i \in \{0,1\}$. After Step 4 of Algorithm 8, it is

$$P \left( \frac{1}{\sqrt{2}} \sum_{i \in \{0,1\}} |i\rangle_T |i\rangle_T U^i |\psi\rangle_{RA} |0\rangle_{E'F} \right), \quad (A47)$$

where $P \equiv P_{T'E'F \rightarrow T''E'}$. For a fixed unitary $P_{T'E'F \rightarrow T''E'}$ of the max-prover and fixed state $|\psi\rangle_{RA}$ of the min-prover, the acceptance probability is then

$$\left\| \langle \phi^0 |_{T'} p_{T''} \left( \frac{1}{\sqrt{2}} \sum_{i \in \{0,1\}} |i\rangle_T |i\rangle_T U^i |\psi\rangle_{RA} |0\rangle_{E'F} \right) \right\|_2^2$$

$$= \frac{1}{2} \left\| \langle \phi^0 |_{T'} p_{T''} \left( \frac{1}{\sqrt{2}} \sum_{i \in \{0,1\}} |i\rangle_T |i\rangle_T U^i |\psi\rangle_{RA} |0\rangle_{E'F} \right) \right\|_2^2$$

(A48)

In a competing-provers quantum interactive proof, the max-prover is trying to maximize the probability that the verifier accepts, while the min-prover is trying to minimize the acceptance probability. Since the max-prover plays second in this game, the acceptance probability of Algorithm 8 is given by

$$\min_{|\psi\rangle_{RA}} \max_P \frac{1}{2} \left\| \langle \phi^0 |_{T'} p_{T''} \sum_{i \in \{0,1\}} |i\rangle_T T^i |\psi\rangle_{RA} |0\rangle_{E'F} \right\|_2^2.$$  (A49)

Applying the analysis of Theorem 1, it follows that

$$\max_P \frac{1}{2} \left\| \langle \phi^0 |_{T'} p_{T''} \sum_{i \in \{0,1\}} |i\rangle_T T^i |\psi\rangle_{RA} |0\rangle_{E'F} \right\|_2^2$$

$$= \frac{1}{2} \left( 1 + \sqrt{F} (N_{A \rightarrow B} (|\psi_{RA} \rangle), N_{A \rightarrow B} (|\psi_{RA} \rangle)) \right). \quad (A50)$$

Thus, after applying the minimization over every input state $|\psi_{RA}\rangle$, the claim in (62) follows. ■

4. Proof of Theorem 4

Proof of Theorem 4. After Step 1 of Algorithm 9, the global state is

$$|\Phi\rangle_{T'T'} |\psi\rangle_{RA} |0\rangle_{E''}. \quad (A51)$$

where we have employed the shorthand $E'' = E'_1 \cdots E'_n$. After Step 6 of Algorithm 9, the global state is

$$|\psi (S_i^{n-1}) \rangle = \frac{1}{\sqrt{2}} \sum_{i \in \{0,1\}} |i\rangle_{T'} |T U^{i,n-1} \prod_{j=1}^{n-1} (S_j^{1,i,j}) |\psi\rangle_{RA} |0\rangle_{E''}, \quad (A52)$$

where we have omitted many of the system labels for simplicity. After Step 8 of Algorithm 9, the global state is

$$P |\psi (S_i^{n-1}) \rangle. \quad (A53)$$

For a fixed unitary $P$ of the max-prover and a fixed pure co-strategy $|\psi, \{S_j^{n-1}\}\rangle$ of the min-prover, the acceptance probability is thus

$$\left\| \langle \phi^0 |_{T'} p_{T''} |\psi, \{S_j^{n-1}\}\rangle \right\|_2^2. \quad (A54)$$

In a double-prover quantum interactive proof, the max-prover is trying to maximize the probability that the verifier accepts, while the min-prover is trying to minimize the acceptance probability. Since the max-prover plays second in this game, the acceptance probability of Algorithm 9 is given by

$$\min_{|\psi, \{S_j^{n-1}\}\rangle} \max_P \left\| \langle \phi^0 |_{T'} p_{T''} \sum_{i \in \{0,1\}} |i\rangle_T T^i |\psi, \{S_j^{n-1}\}\rangle \right\|_2^2. \quad (A55)$$

Applying the analysis of Theorem 1, it follows that
max \| \langle \Phi|_{T''T} P|\varphi(\psi, \{S^i\}_{n-1}^n)\rangle \|_2^2
\]
\[= \frac{1}{2} \left( 1 + \sqrt{F(N^0, n \circ S(n-1), N^1, n \circ S(n-1))} \right). \]

(A56)

Thus, after applying a minimization over every pure co-strategy $S(n-1)$, the claim in (77) follows. \[\blacksquare\]

5. Proof of Theorem 5

Proof of Theorem 5. After Step 2 of Algorithm 10, the global state is

\[|\Phi\rangle_{T''T}\psi_{RA}|0\rangle_{E'}. \]

(A57)

After Step 3, the global state is

\[\frac{1}{\sqrt{2}} \sum_{i \in \{0,1\}} |i\rangle_{T'} |i\rangle_{T'} U_{AE'\rightarrow BE}^i |\psi\rangle_{RA}|0\rangle_{E'}. \]

(A58)

After Step 5, it is

\[\frac{1}{\sqrt{2}} P \sum_{i \in \{0,1\}} |i\rangle_{T'} |i\rangle_{T'} U_{AE'\rightarrow BE}^i |\psi\rangle_{RA}|0\rangle_{E'}. \]

(A59)

where $P \equiv P_{T'EF \rightarrow T''E'}$. For a fixed state $|\psi\rangle_{RA}$ and unitary $P_{T'EF \rightarrow T''E'}$ of the prover, the acceptance probability is

\[\frac{1}{2} \| \langle \Phi|_{T''T} P \sum_{i \in \{0,1\}} |i\rangle_{T'} |i\rangle_{T'} U_{AE'\rightarrow BE}^i |\psi\rangle_{RA}|0\rangle_{E'} \|_2^2. \]

(A60)

In a QIP algorithm, the prover chooses his actions in order to maximize the acceptance probability, so that the acceptance probability is

\[\frac{1}{2} \sup_{|\psi\rangle_{RA}} \left\| \langle \Phi|_{T''T} P \sum_{i \in \{0,1\}} |i\rangle_{T'} |i\rangle_{T'} U_{AE'\rightarrow BE}^i |\psi\rangle_{RA}|0\rangle_{E'} \right\|_2^2. \]

(A61)

By the same reasoning employed in the proof of Theorem 1, we conclude that

\[\frac{1}{2} \sup_P \left\| \langle \Phi|_{T''T} P \sum_{i \in \{0,1\}} |i\rangle_{T'} |i\rangle_{T'} U_{AE'\rightarrow BE}^i |\psi\rangle_{RA}|0\rangle_{E'} \right\|_2^2
\]
\[= \frac{1}{2} \left( 1 + \sqrt{F(N^0, A \circ B(\rho_A), N^1, A \circ B(\rho_A))} \right). \]

(A62)

where $\rho_A$ is the reduced state of $\psi_{RA}$ (i.e., $Tr_R|\psi_{RA}\rangle = \rho_A$). Now including the optimization over every pure state $\psi_{RA}$, we conclude the claim in (82). \[\blacksquare\]

6. Proof of Theorem 6

Proof of Theorem 6. After Step 2 of Algorithm 11, the global state is

\[\sum_{x \in \mathcal{X}} \sqrt{p(x)} |x\rangle_{xT} |\psi^x\rangle_{RS}. \]

(A63)

After Step 4, it is

\[P \sum_{x \in \mathcal{X}} \sqrt{p(x)} |x\rangle_{xT} |\psi^x\rangle_{RS}|0\rangle_{F}, \]

(A64)

where $P \equiv P_{T'RF \rightarrow T''F'}$. Then, for a fixed unitary $P_{T'RF \rightarrow T''F'}$, the acceptance probability is

\[\left\| \langle \Phi|_{T''T} P \sum_{x \in \mathcal{X}} \sqrt{p(x)} |x\rangle_{xT} |\psi^x\rangle_{RS}|0\rangle_{F} \right\|_2^2 = \sup_{|\phi\rangle_{F'S}} \left\| \langle \Phi|_{T''T} \langle \phi|_{F'S} P \sum_{x \in \mathcal{X}} \sqrt{p(x)} |x\rangle_{xT} |\psi^x\rangle_{RS}|0\rangle_{F} \right\|_2^2, \]

(A65)

where the optimization is over every pure state $|\phi\rangle_{F'S}$ and we have used the fact $\|\psi\|_2^2 = \sup_{|\phi\rangle_{F'S}} \|\langle \phi|_{F'S}\|_2^2 = 1$. This implies that the acceptance probability is given by

\[\sup_{|\phi\rangle_{F'S}} \left\| \langle \Phi|_{T''T} \langle \phi|_{F'S} P \sum_{x \in \mathcal{X}} \sqrt{p(x)} |x\rangle_{xT} |\psi^x\rangle_{RS}|0\rangle_{F} \right\|_2^2. \]

(A66)

Recall Uhlmann’s theorem [Uhl76], which is the statement that

\[F(\omega_C, \tau_C) = \sup_{V_B} |\langle \varphi^*|_{BC} V_B \otimes I_C |\varphi^*|_{BC}\rangle|^2, \]

(A67)

where $\omega_C$ and $\tau_C$ are density operators with respect to purifications $|\varphi^*\rangle_{BC}$ and $|\varphi^\tau\rangle_{BC}$ and the optimization is over every unitary $V_B$. Observing that the unitary $P_{T'RF \rightarrow T''F'}$ acts on systems $T'RF$ of $\sum_{x \in \mathcal{X}} \sqrt{p(x)} |x\rangle_{xT} |\psi^x\rangle_{RS}|0\rangle_{F}$ and systems $T''F'$ of $|\Phi|_{T''T} |\varphi\rangle_{F'S}$, that their respective reduced states on systems $TS$ are

\[\sum_{x \in \mathcal{X}} p(x) |x\rangle_{xT} \otimes \rho_S^x, \]

(A68)

\[\pi_T \otimes \sigma_S, \]

(A69)

where $\pi_T$ is the maximally mixed state and $\sigma_S := Tr_F |\varphi\rangle_{F'S}$, and applying Uhlmann’s theorem, we conclude that the acceptance probability is given by

\[\sup_{\sigma_S} F \left( \sum_{x \in \mathcal{X}} p(x) |x\rangle_{xT} \otimes \rho_S^x, \pi_T \otimes \sigma_S \right) \]

(A70)

\[= \left[ \sup_{\sigma_S} \sqrt{F} \left( \sum_{x \in \mathcal{X}} p(x) |x\rangle_{xT} \otimes \rho_S^x, \pi_T \otimes \sigma_S \right) \right]^2 \]

(A71)
as a consequence of Theorem 6. Thus, we arrive at the claim in (98) by minimizing over every state $\psi_{RA}$ of the min-prover.

The upper bound in (99) follows from the upper bound in (91). Indeed, for a fixed state $\psi_{RA}$ of the min-prover, the acceptance probability in (A78) is bounded from above by

$$\frac{1}{d} + \frac{2}{d} \sum_{x,y \in X; x < y} \sqrt{p(x)p(y)}\sqrt{F(N_{A \to B}^x(\psi_{RA}), N_{A \to B}^y(\psi_{RA}))}.$$ 

After taking infima, we arrive at (99).

The final statement follows from the same reasoning employed at the end of the proof of Theorem 6. ■

Appendix B: Number of samples for Fidelity-Pure-Pure

In Theorem 12, we argued that the problem Fidelity-Pure-Pure is BQP-complete; i.e., every problem in BQP can be reduced to this problem in polynomial time. In this section, we discuss the number of samples required to obtain a desired accuracy and confidence. Let us first recall Hoeffding’s bound.

**Lemma 18 (Hoeffding’s Bound)** Suppose that we are given $n$ independent samples $Y_1, \ldots, Y_n$ of a bounded random variable $Y$ taking values in the interval $[a, b]$ and having mean $\mu$. Set

$$Y_n := \frac{1}{n} (Y_1 + \ldots + Y_n)$$ 

(B1)

as the sample mean. Let $\epsilon \in (0, 1)$ be the desired accuracy, and let $1 - \delta$ be the desired success probability, where $\delta \in (0, 1)$. Then

$$\text{Pr} \{ |\hat{Y}_n - \mu| \leq \epsilon \} \geq 1 - \delta,$$

(B2)

as long as

$$n \geq \frac{M^2}{2\epsilon^2} \ln \left( \frac{2}{\delta} \right),$$

(B3)

where $M := b - a$.

In the main text, we mapped a general BQP algorithm to Fidelity-Pure-Pure. In a general BQP algorithm, we measure a single qubit called the decision qubit, leading to a random variable $Y$ taking the value 0 with probability $1 - p$ and the value 1 with probability $p$, where $p$ is the acceptance probability of the algorithm. We repeat this procedure $n$ times and label the outcomes $Y_1, \ldots, Y_n$. We output the mean

$$\hat{Y}_n = \frac{1}{n} (Y_1 + \ldots + Y_n)$$

(B4)

as an estimate for the true value $p$ (as seen in (150))

$$p = \langle x | s \rangle_{A} Q^\dagger (|1\rangle_{D} \otimes I_G) Q |x\rangle_{S} |0\rangle_{A}.$$ 

(B5)
By plugging into Lemma 18, setting

$$\mu = p$$  \hspace{1cm} \text{(B6)}

therein, and taking \(n\) to satisfy the condition \(n \geq \frac{1}{\varepsilon^2} \ln \left(\frac{2}{\delta}\right)\), we can achieve an error \(\varepsilon\) and confidence \(\delta\) (as defined in (B2)).

Now, we see from (153) that the modified algorithm has an acceptance probability \(p^2\), i.e., equal to the square of the original BQP problem’s acceptance probability. In the modified algorithm, we measure the decision qubit, leading to a random variable \(Z\) taking value 0 with probability \(1 - p^2\) and the value 1 with probability \(p^2\). We repeat the procedure \(m\) times and label the outcomes \(Z_1, \ldots, Z_m\). We output the mean

$$Z_m = \frac{1}{m} (Z_1 + \ldots + Z_m)$$  \hspace{1cm} \text{(B7)}

as an estimate for the true value \(p^2\) (as seen in (153)). Setting \(\hat{\mu} = p^2\), and plugging into Lemma 18, it follows that

$$\Pr[|Z_m - \hat{\mu}| \leq \varepsilon^2] \geq 1 - \delta,$$  \hspace{1cm} \text{(B8)}

if

$$m \geq \frac{1}{2\varepsilon^4} \ln \left(\frac{2}{\delta}\right).$$  \hspace{1cm} \text{(B9)}

Consider the following inequalities:

$$\varepsilon^2 \geq |Z_m - \hat{\mu}|$$
$$= |Z_m - \mu^2|$$
$$= |\sqrt{Z_m} - \mu| |\sqrt{Z_m} + \mu|$$
$$\geq |\sqrt{Z_m} - \mu|^2,$$  \hspace{1cm} \text{(B10)}

where the second inequality is derived from the fact that \(Z_m, \mu \in [0, 1]\), so that \(|Z_m + \mu| \geq |Z_m - \mu|\). Thus,

$$|\sqrt{Z_m} - \mu| \leq \varepsilon.$$  \hspace{1cm} \text{(B11)}

In other words,

$$\varepsilon^2 \geq |Z_m - \mu^2| \implies \varepsilon \geq |\sqrt{Z_m} - \mu|$$  \hspace{1cm} \text{(B12)}

so that

$$\Pr \left[ |\sqrt{Z_m} - \mu| \leq \varepsilon \right] \geq \Pr[|Z_m - \mu^2| \leq \varepsilon^2]$$
$$\geq 1 - \delta.$$  \hspace{1cm} \text{(B13)}

Thus, \(\sqrt{Z_m}\) is an estimator for \(p\) and taking

$$m \geq \frac{1}{2\varepsilon^4} \ln \left(\frac{2}{\delta}\right)$$  \hspace{1cm} \text{(B14)}

suffices to achieve an error \(\varepsilon\) and confidence \(\delta\) in estimating \(p\).

### Appendix C: Approximate fixed points and Deutschian closed timelike curves

The computational complexity of computation assisted by Deutschian closed timelike curves (CTCs) was solved in [AW09], in which these authors showed that the power of classical and quantum computing are equivalent and equal to PSPACE, which is the class of decision problems solvable with polynomial memory. Let us briefly review these results. In the Deutschian model of CTCs [Deu01], we suppose that chronology-respecting qubits in a state \(\rho\) can interact with chronology-violating qubits in a state \(\sigma\) according to a unitary transformation \(U\). Let \(S\) denote the quantum system for the chronology-respecting qubits, and let \(C\) denote the quantum system for the chronology-violating qubits. Then the output state of the transformation is as follows:

$$\text{Tr}_C[U_S \rho \otimes \sigma_C U_S^\dagger].$$  \hspace{1cm} \text{(C1)}

In an effort to avoid grandfather and unproved theorem paradoxes, Deutsch postulates that nature imposes the following self-consistency condition on the state of the CTC qubits:

$$\sigma_C = \text{Tr}_S[U_S \rho \otimes \omega_C U_S^\dagger].$$  \hspace{1cm} \text{(C2)}

At a first glance, this condition might seem innocuous, but its implications for quantum information processing are dramatic, essentially due to the fact that (C2) allows for non-linear evolutions, which are disallowed in standard quantum mechanics. Indeed, quantum processors assisted by Deutschian CTCs can violate the uncertainty principle [BHW09, PRM13], can break the no-cloning theorem [BWW13, YAT15], and can solve computational problems believed to be difficult [AW09].

The connection of Deutschian CTCs (D-CTCs) with fixed points of channels is that the condition in (C2) demands that the state of the CTC system be a fixed point of the quantum channel \(\mathcal{N}_{U,\rho}\):

$$\omega_C \rightarrow \mathcal{N}_{U,\rho}(\omega_C) := \text{Tr}_S[U_S \rho \otimes \omega_C U_S^\dagger].$$  \hspace{1cm} \text{(C3)}

Thus, this is how Section VI connects with Deutschian CTCs.

The class of computational problems efficiently decidable by a quantum computer assisted by D-CTCs is called BQP_{CTC}, and it is formally defined as follows. Set \(\delta \in (0, 1/2)\). Let \(G\) be a universal set of quantum gates. A quantum D-CTC algorithm is a deterministic polynomial time algorithm that takes as input a string \(x \in \{0, 1\}^n\) and produces an encoding of a unitary quantum circuit \(U\) using gates from \(G\). This unitary acts on two systems of qubits, called \(S\) and \(C\) as discussed above, which consist of \(p(n)\) and \(q(n)\) qubits, respectively, where \(p(n)\) and \(q(n)\) are polynomials. The system \(S\) is initialized to the all-zeros state \(|0\rangle^\otimes_{p(n)}\), and the system \(C\) is initialized to a state \(\sigma_C\) satisfying the causal
Let $M$ be a measurement of the last qubit of $S$ in the computational basis. The algorithm accepts the input $x$ if

$$M(\text{Tr}_C[U_{SC}(|0\rangle|0\rangle_{S}^{\otimes p(n)} \otimes \sigma_C) U_{SC}^{\dagger}])$$

results in the output 1 with probability at least $1 - \delta$ for every state $\sigma_C$ satisfying (C2). The algorithm rejects if (C5) results in output 1 with probability no larger than $\delta$ for every state $\sigma_C$ satisfying (C2). The algorithm decides the promise problem $A = A_{yes} \cup A_{no} \subseteq \{0, 1\}^*$ (where $A_{yes} \cap A_{no} = \emptyset$) if the algorithm accepts every input $x \in A_{yes}$ and rejects every input $x \in A_{no}$. BQP$_{CTC}$ is the class of all promise problems that are decided by some quantum D-CTC algorithm.

It is already known from [AW09] that BQP$_{CTC} = \text{PSPACE}$, and it is also known that QIP = PSPACE [JJUW11]. Thus, it immediately follows from these results that BQP$_{CTC} = \text{QIP}$. Here, we discuss an attempt at a direct proof that BQP$_{CTC} \subseteq \text{QIP}$, which ideally would be arguably simpler to see than by examining the proofs of the equalities BQP$_{CTC} = \text{PSPACE}$ and QIP = PSPACE individually. However, there are some difficulties in establishing this direct proof. We note here that this is related to an open question posed in [Aar05, Section 8], the spirit of which is to find a direct proof of the containment BQP$_{CTC} \subseteq \text{QIP}$.

Consider the following purported algorithm for simulating BQP$_{CTC}$ in QIP:

**Algorithm 20** The algorithm proceeds as follows:

1. The verifier prepares a state

$$|\Phi\rangle_{T+T} := \sum_{\ell=0}^{L-1} \sqrt{\frac{1}{L}} |\ell\rangle_{T+T}$$

on registers $T'$ and $T$ and prepares system $S_{T^{L}}$ in the all-zeros state $|0\rangle_{S_{T^{L}}}$.

2. The prover transmits the system $C$ of the state $|\psi\rangle_{RC}$ to the verifier.

3. Using the circuit $U_{SC}$, the verifier performs the following controlled unitary:

$$\sum_{\ell=0}^{L-1} |\ell\rangle_{T} \otimes U_{S_{T^{L}}}^{\ell}$$

where

$$U_{S_{T^{L}}}^{\ell} := (U_{s_{2}C} \circ \ldots \circ U_{s_{1}C} \circ s_{1}C)_{\ell \text{ times}}$$

4. The verifier transmits systems $T'$ and $S_{T^{L}}$ to the max-prover.

5. The prover prepares a system $F$ in the $|0\rangle_F$ state and acts on systems $T''$, $S^{L'}$, and $F$ with a unitary $P_{T''S^{L'}F} = T''F$, to produce the output systems $T''$ and $F'$, where $T''$ is a qudit system.

6. The prover sends system $T''$ to the verifier, who then performs a Bell measurement

$$\{\Phi_{T''T}, I_{T''T} - \Phi_{T''T}\}$$

on systems $T''$ and $T$, where $\Phi_{T''T}$ is defined in (88). The verifier then initializes a system $S_{L+1}$ to the all-zeros state $|0\rangle_{S_{L+1}}$, performs the unitary $U_{S_{L+1}}C$, and measures the decision qubit of system $S_{L+1}$. The verifier accepts if and only if the outcome $\Phi_{T''T}$ occurs and the decision qubit is measured to be in the $|1\rangle$ state.

**Proposition 2** The acceptance probability of Algorithm 20 is equal to

$$\left[\sup_{\rho_C, \sigma_G} \frac{1}{L} \sum_{\ell=0}^{L-1} \sqrt{\text{Tr}(|1\rangle_{D} \otimes \sigma_G, U_{SC}(|0\rangle_{S} \otimes \mathcal{N}^{\ell}(|\rho_C\rangle) U_{SC}^{\dagger})}\right]^2,$$

where

$$\mathcal{N}(\omega_C) := \text{Tr}_S[U_{SC}(|0\rangle_{S} \otimes \omega_C) U_{SC}^{\dagger}].$$

**Proof.** This follows by employing reasoning similar to that for [Ros09, Lemma 4.2] (see also [KW00]). This reasoning is also very similar to the reasoning used around (A65)–(A72). For a fixed state $|\psi\rangle_{RC}$ of the prover, the global state after Step 6 of Algorithm 20, but before the measurements, is

$$P_{U_{S_{L+1}}C} \sum_{\ell=0}^{L-1} \sqrt{\frac{1}{L}} |\ell\rangle_{T} T' T'' U_{S_{T^{L}}}^{\ell} |0\rangle_{S_{L+1}} |\psi\rangle_{RC},$$

where $P \equiv P_{T''S^{L'}F} = T''F$. Then, by splitting the systems $S_{L+1}C$ into the decision qubit $D$ and denoting all other qubits by $G$, the acceptance probability is given by

$$\left\| \sum_{\ell=0}^{L-1} \sqrt{\frac{1}{L}} |\ell\rangle_{T} T' T'' U_{S_{T^{L}}}^{\ell} |0\rangle_{S_{L+1}} |\psi\rangle_{RC} \right\|_2^2 = \sup_{|\varphi\rangle_{F'}} \left\| \sum_{\ell=0}^{L-1} \sqrt{\frac{1}{L}} |\ell\rangle_{T} T' T'' U_{S_{T^{L}}}^{\ell} |0\rangle_{S_{L+1}} |\psi\rangle_{RC} \right\|_2^2.$$

Considering that the reduced state of

$$\sum_{\ell=0}^{L-1} \sqrt{\frac{1}{L}} |\ell\rangle_{T} T' T'' U_{S_{T^{L}}}^{\ell} |0\rangle_{S_{L+1}} |\psi\rangle_{RC},$$

after tracing over all systems sent to the prover, is

$$\frac{1}{L} \sum_{\ell=0}^{L-1} |\ell\rangle_{T} \otimes \mathcal{N}^{\ell}(|\rho_C\rangle),$$

the acceptance probability is

$$\left[\sup_{\rho_C, \sigma_G} \frac{1}{L} \sum_{\ell=0}^{L-1} \sqrt{\text{Tr}(|1\rangle_{D} \otimes \sigma_G, U_{SC}(|0\rangle_{S} \otimes \mathcal{N}^{\ell}(|\rho_C\rangle) U_{SC}^{\dagger})}\right]^2.$$
where $\rho_C := \Tr[\psi_{RC}]$, and the reduced state of $|\Phi\rangle_{T\rightarrow T}|1_D\rangle|\psi\rangle_{FG}$, after tracing over all systems not transmitted by the prover, is

$$\frac{1}{L} \sum_{\ell=0}^{L-1} |\ell\rangle \langle \ell |_T \otimes |1\rangle |D \otimes \sigma_G,$$  \hspace{1cm} (C15)

where $\sigma_G := \Tr_{F'}[\varphi_{F'G}]$, we conclude by Uhlmann’s theorem that (C13) is equal to

$$\sup_{\sigma_G} F\left(\frac{1}{L} \sum_{\ell=0}^{L-1} |\ell\rangle \langle \ell |_T \otimes N^\ell(\rho_C), \frac{1}{L} \sum_{\ell=0}^{L-1} |\ell\rangle \langle \ell |_T \otimes |1\rangle |D \otimes \sigma_G \right) = \left[ \sup_{\sigma_G} \frac{1}{L} \sum_{\ell=0}^{L-1} \sqrt{F\left(U_{SC}(|0\rangle \langle 0 |_S \otimes N^\ell(\rho_C))U_{SC}^\dagger \right)} \right]^2. \hspace{1cm} (C16)$$

We conclude the expression in the statement of the theorem after optimizing over all input states of the prover.

In order to establish that $\text{BQP}_{\text{CTC}}$ is contained in QIP, it is necessary to map yes-instances of the former to yes-instances of the latter, and the same for the no-instances. Accomplishing the first part of the task is straightforward. A yes-instance of $\text{BQP}_{\text{CTC}}$ implies that there exists a fixed-point state $\rho_C$ such that

$$\Tr\left([|1\rangle |1 \otimes I_G]U_{SC}([0\rangle \langle 0 |_S \otimes \rho_C)U_{SC}^\dagger \right) \geq 1 - \delta. \hspace{1cm} (C17)$$

Thus, the prover transmits such a fixed-point state $\rho_C$ to the verifier, and we find that the acceptance probability is not smaller than

$$\left[ \sup_{\sigma_G} \frac{1}{L} \sum_{\ell=0}^{L-1} \sqrt{F\left(U_{SC}(|0\rangle \langle 0 |_S \otimes N^\ell(\rho_C))U_{SC}^\dagger \right)} \right]^2 \geq \sup_{\sigma_G} F\left(|1\rangle |1 \otimes \sigma_G, U_{SC}([0\rangle \langle 0 |_S \otimes \rho_C)U_{SC}^\dagger \right) \geq \sup_{\sigma_G} \frac{1}{L} \sum_{\ell=0}^{L-1} \sqrt{F\left(U_{SC}(|0\rangle \langle 0 |_S \otimes N^\ell(\rho_C))U_{SC}^\dagger \right)} \geq \sup_{\sigma_G} \frac{1}{L} \sum_{\ell=0}^{L-1} \sqrt{\langle\varphi|F\cdot G|\psi\rangle U_{SC}|\psi\rangle_{RC}^2} \geq \frac{1}{L} \sum_{\ell=0}^{L-1} N^\ell(\omega_C).$$

The first inequality follows because $N^\ell(\rho_C) = \rho_C$ for all $\ell$.

It is less clear how to handle the case of a no-instance of $\text{BQP}_{\text{CTC}}$, because the definition of this complexity class only specifies the behavior of the circuit when $\rho_C$ is an exact fixed point of $N$. Algorithm 20 attempts to verify whether the prover sends a fixed point, but it only actually verifies whether the prover sends a state that is an approximate fixed point. The acceptance probability of Algorithm 20 is given by (C10) and is bounded from above by

$$\sup_{\rho_C, \sigma_G} F\left([|1\rangle |1 \otimes \sigma_G, U_{SC}([0\rangle \langle 0 |_S \otimes N_L(\rho_C))U_{SC}^\dagger \right) \leq \sup_{\rho_C} \sup_{\sigma_G} \frac{1}{L} \sum_{\ell=0}^{L-1} \sqrt{F\left(U_{SC}(|0\rangle \langle 0 |_S \otimes \rho_C)U_{SC}^\dagger \right)} \geq \sup_{\rho_C} \frac{1}{L} \sum_{\ell=0}^{L-1} N^\ell(\omega_C).$$

where the bounds follow from concavity of root fidelity and the data-processing inequality for fidelity. In the above, $N_L$ is the Cesaro mean channel:

$$N_L(\omega_C) := \frac{1}{L} \sum_{\ell=0}^{L-1} N^\ell(\omega_C).$$

This channel has the property that the sequence $\{N_L\}_L$ converges to the fixed-point projection channel $P := \lim_{L \rightarrow \infty} N_L$ of $N$, so that $P(\omega_C)$ is guaranteed to be a fixed point of $N$ for every input state $\omega_C$ [Wol12]. It is not clear how to obtain a channel independent bound that relates the convergence of $N_L$ to $P$, as a function of $L$ alone. Furthermore, it is likely not possible that the closeness of $N_L$ to $P$ could generally be inverse polynomial in $L$; for if it were, then one could simulate $\text{BQP}_{\text{CTC}}$ in $\text{BQP}$, because the verifier could apply the map $N_L$ and generate a fixed point of the channel without the help of the prover. However, we now know that $\text{BQP}_{\text{CTC}} = \text{PSPACE}$ [AW09], and it is widely believed that PSPACE $\neq$ BQP. In the case of a no-instance of $\text{BQP}_{\text{CTC}}$, it is thus not clear how to relate the acceptance probability of Algorithm 20 to the acceptance probability of the no-instance of $\text{BQP}_{\text{CTC}}$. We leave this as a curious open question.