A smallest computable entanglement monotone

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Abstract—The Rains relative entropy of a bipartite quantum state is the tightest known upper bound on its distillable entanglement – which has a crisp physical interpretation of entanglement as a resource – and it is efficiently computable by convex programming. It has not been known to be a selective entanglement monotone in its own right. In this work, we strengthen the interpretation of the Rains relative entropy by showing that it is monotone under the action of selective operations that completely preserve the positivity of the partial transpose, reasonably quantifying entanglement. That is, we prove that Rains relative entropy of an ensemble generated by such an operation does not exceed the Rains relative entropy of the initial state in expectation, giving rise to the smallest, most conservative known computable selective entanglement monotone. Additionally, we show that this is true not only for the original Rains relative entropy, but also for Rains relative entropies derived from various Rényi relative entropies. As an application of these findings, we prove, in both the non-asymptotic and asymptotic settings, that the probabilistic approximate distillable entanglement of a state is bounded from above by various Rains relative entropies.

I. INTRODUCTION

Entanglement is the feature at the heart of quantum mechanics. One can go as far to say it is the defining property that distinguishes it from a classical theory [11]. In quantum information, entanglement is seen as a resource for information-processing tasks [2], [3]. For this reason, substantial efforts have been made to quantify and capture it, giving rise to a comprehensive theory of entanglement [2], [3].

Many measures of entanglement have been presented and their properties studied over the years. Among those, one of the most prominent ones is the distillable entanglement [4], [5], [6], which captures entanglement as a resource: It quantifies the rate at which one can meaningfully extract Bell states, which can then be used, e.g., in quantum key distribution [7]. Physically meaningful as this quantity is, it is notoriously difficult to compute and known to be not convex [8]. It may not even be Turing computable [8], [9], [10].

Since the distillable entanglement is so important conceptually, and also because a more pragmatic mindset has become more common over the years, researchers have resorted to using upper bounds on the distillable entanglement. The tightest known upper bound is the Rains relative entropy [11], a quantity that interpolates the relative entropy of entanglement [12] and the logarithmic negativity [13], [14], [15]. Due to this finding and since it can be computed as a convex program [16], [17], the Rains relative entropy is considered a practically important bound capturing notions of entanglement.

It has been well known since the original work [11] that the Rains relative entropy is non-increasing under the action of trace-preserving channels that completely preserve the positivity of the partial transpose (PPT), and as such, it is non-increasing under physically well motivated local operations and classical communication (LOCC) channels that reflect the separated laboratory paradigm for bipartite quantum systems. This is in fact one of the properties required to prove that it is an upper bound on distillable entanglement. That said, importantly, what has been left open over the years is to determine whether it actually has the stronger property of being a monotone under selective quantum operations: This property to be a selective entanglement monotone [18], [19] is, however, an important property so that the Rains quantity can be regarded as a quantity meaningfully capturing entanglement according to this stronger notion. This means that the average Rains relative entropy of an ensemble produced by suitable selective operations does not exceed that of the original state.

In this work, we bring this problem to rest by proving that the Rains relative entropy is a selective PPT monotone (thus implying that it is also a selective LOCC monotone). Since it is the tightest known upper bound on the distillable entanglement – which itself is the smallest reasonable entanglement monotone [19] – it is at the same time established as the smallest known computable measure of entanglement. Hence, it reasonably quantifies entanglement in a most conservative fashion, and as such provides a good guideline to assess quantum correlations in practically relevant settings. Our proof has a general form, and so it applies not only to the original Rains relative entropy, but also to others constructed from various Rényi relative entropies.

As an application, we prove that the Rains relative entropy is an upper bound on the probabilistic approximate distillable entanglement of a bipartite state, in both the non-asymptotic and asymptotic settings. Since the probabilistic approximate distillable entanglement of a state is an upper bound on the usual distillable entanglement and it arguably has a more direct connection to experimental practice, our result strengthens the interpretation of the Rains relative entropy, thus giving the tightest known upper bound on the probabilistic approximate distillable entanglement of a state.
II. Preliminaries

In this preliminary section, we set notation and review background material. In particular, we recall the definitions of various relative entropies and their properties. These are then used to define Rains relative entropies, which are the signature entanglement measures considered in this work. After that, we present some properties of the Rains relative entropies and recall the definition of the set of completely positive-partial-transpose (PPT) preserving channels.

A. Quantum states, channels, and relative entropies

In what follows, naturally, notions of quantum states, channels and relative entropies feature strongly; we point to the textbooks [20, 21, 22, 23, 24] for more background on quantum information theory. We denote with $S(\mathcal{H})$ the set of quantum states (unit-trace, positive semi-definite operators) acting on a Hilbert space $\mathcal{H}$. Let $\mathcal{L}(\mathcal{H})$ denote the set of linear operators acting on $\mathcal{H}$, and let $\mathcal{L}_+(\mathcal{H})$ denote the set of positive semi-definite operators acting on $\mathcal{H}$. The following data-processing inequalities[26], [27]. The following limit holds[26], [27]

$$\lim_{\alpha \to 1} D_\alpha(\omega|\tau) = D(\omega|\tau).$$

The following property of the quantum relative entropy is defined for $H$ operators acting on $\mathcal{H}$.

$$D(\omega|\tau) := \sum \omega(x) \log_2 \frac{\omega(x)}{\tau(x)}$$

where $(F1) \equiv \text{supp}(\omega) \subseteq \text{supp}(\tau)$. The sandwiched Rényi relative entropy is defined for $\alpha \in (0, 1) \cup (1, \infty)$ as

$$\tilde{D}_\alpha(\omega|\tau) := \frac{1}{\alpha-1} \log_2 \tilde{Q}_\alpha(\omega|\tau),$$

$$\tilde{Q}_\alpha(\omega|\tau) := \left\{ \begin{array}{ll}
\text{Tr}[\omega^{1-\alpha} \tau^{\alpha-1} & \text{if (F1)} \\
\infty & \text{else,}
\end{array} \right.\tag{1}$$

where $(F2) \equiv \alpha \in (0, 1) \lor (\alpha > 1 \land \text{supp}(\omega) \subseteq \text{supp}(\tau))$ [26], [27]. The following limit holds [26], [27]

$$\lim_{\alpha \to 1} \tilde{D}_\alpha(\omega|\tau) = D(\omega|\tau).$$

The following data-processing inequalities

$$D(\omega|\tau) \geq D(N(\omega)|N(\tau)), \tilde{D}_\alpha(\omega|\tau) \geq \tilde{D}_\alpha(N(\omega)|N(\tau)),\tag{3}$$

hold for $\omega \in S(\mathcal{H})$, $\tau \in \mathcal{L}_+(\mathcal{H})$, $N \in \text{CPPT}(\mathcal{H}, \mathcal{H}')$, and $\alpha \in [1/2, 1) \cup (1, \infty)$, with the first inequality established by ref. [23] and the second by ref. [29] (see also ref. [30]). The Petz and geometric Rényi relative entropies are defined by the same construction in (1), but by replacing $Q_\alpha$ with $Q_\alpha(\omega|\tau) := \text{Tr}[\omega^{\alpha-1}\tau^{1-\alpha}]$ [31] and $\tilde{Q}_\alpha(\omega|\tau) := \text{Tr}[\tau^{\alpha-1/2}\omega^{1/2\alpha}]$ [32], [33], [24], respectively. A data-processing inequality, similar to the above, holds for them for $\alpha \in (0, 1) \cup (1, 2]$. The following property of the quantum relative entropy is well known [24].

Lemma 1 (Direct-sum property). Let $\kappa \in S(\mathcal{H}_{X_A})$ and $\lambda \in \mathcal{L}_+(\mathcal{H}_{X_A})$ be classical–quantum, i.e., of the form

$$\kappa := \sum_x p(x)|x\rangle \langle x| \otimes \kappa_x, \quad \lambda := \sum_x q(x)|x\rangle \langle x| \otimes \lambda_x,$$

where $\{|x\rangle\}^x$ is an orthonormal basis for $\mathcal{H}_X$, $\{p(x)\}^x$ is a probability distribution, $\kappa_x \in S(\mathcal{H}_A)$ for all $x$, $\{q(x)\}^x$ is a non-negative function, and $\lambda_x \in \mathcal{L}_+(\mathcal{H}_A)$ for all $x$. Then

$$D(\kappa||\lambda) = D(p||q) + \sum_x p(x)D(\kappa_x||\lambda_x),$$

where the classical relative entropy is defined as

$$D(p||q) := \left\{ \begin{array}{ll}
\sum_x p(x) \log_2 \left( \frac{p(x)}{q(x)} \right) & \text{if supp}(p) \subseteq \text{supp}(q) \\
+\infty & \text{else.}
\end{array} \right.\tag{4}$$

The same inequality holds for $\alpha > 1$ with $D_\alpha$ replaced by the Petz or geometric Rényi relative entropy.

B. Rains relative entropies

We now turn to discussing the central quantity of this work. Let $\rho \in S(\mathcal{H}_{AB})$ denote a bipartite state acting on a tensor-product Hilbert space $\mathcal{H}_{AB} := \mathcal{H}_A \otimes \mathcal{H}_B$. To define it, we need to refer to the partial transpose of a quantum state. To define it, denote with $T() := \sum_i |i\rangle \langle j| |i\rangle \langle j|$ the transpose map, with $\{|i\rangle\}$ an orthonormal basis for $\mathcal{H}_B$, so that $\text{id} \otimes T : \mathcal{L}(\mathcal{H}_{AB}) \rightarrow \mathcal{L}(\mathcal{H}_{AB})$ is the partial transpose map, given by $(\text{id} \otimes T)(\cdot) := \sum_i |i\rangle \langle i| \otimes |j\rangle \langle j|$. The Rains relative entropy has originally been defined as [35], [11]

$$R(\rho) := \inf_{\sigma \in S(\mathcal{H}_{AB})} \left( D(\rho||\sigma) + \log_2 \| (\text{id} \otimes T)(\sigma) \|_1 \right),\tag{5}$$

where $\| \cdot \|_1$ denotes the trace norm. This expression has later been identified to be equal to the convex program [36]

$$R(\rho) = \inf_{\sigma \in \text{PPT}(\mathcal{H}_{AB})} D(\rho||\sigma),\tag{6}$$

$$\text{PPT}^+(\mathcal{H}_{AB}) := \{ \sigma : \sigma \in \mathcal{L}_+(\mathcal{H}_{AB}) \land \| (\text{id} \otimes T)(\sigma) \|_1 \leq 1 \}.$$

For $\alpha \in (0, 1) \cup (1, \infty)$, we define with

$$\tilde{R}_\alpha(\rho) := \inf_{\sigma \in \text{PPT}(\mathcal{H}_{AB})} \tilde{D}_\alpha(\rho||\sigma)\tag{7}$$

the sandwiched Rains relative entropy, satisfying [37]

$$\lim_{\alpha \to 1} \tilde{R}_\alpha(\rho) = R(\rho).\tag{8}$$
By replacing $\tilde{D}_{\alpha}$ in (7) with either the Petz or geometric Rényi relative entropy, we can construct alternative Rains relative entropies [37], [24]. They are subadditive in that
\[
R(\rho_0 \otimes \rho_1) \leq R(\rho_0) + R(\rho_1), \quad \tilde{R}_\alpha(\rho_0 \otimes \rho_1) \leq \tilde{R}_\alpha(\rho_0) + \tilde{R}_\alpha(\rho_1),
\]
for all $\alpha \in (0, 1) \cup (1, \infty)$, where $\rho_i \in S(\mathcal{H}_{A_iB_i})$ for $i \in \{0, 1\}$ and the bipartite cut for $\rho_0 \otimes \rho_1$ is $A_0A_1|B_0B_1$. As a direct consequence of the joint convexity of quantum relative entropy [35], the Rains relative entropy is convex [11],
\[
R(\rho) \leq \sum_x p(x) R(\rho_x),
\]
where $\bar{\rho} := \sum_x p(x) \rho_x$, with $\{p(x)\}_x$ a probability distribution and $\rho_x \in S(\mathcal{H}_{A})$ for all $x$. Indeed, building on a statement from ref. [36, Lemma 3], one can capture the Rains relative entropy in a more general way, proven in Appendix B.

**Lemma 3** (Scaling property). Let $D(\omega||\tau)$ be a real-valued function of $\omega \in S(\mathcal{H})$ and $\tau \in L_+^+(\mathcal{H})$, which satisfies
\[
D(\omega||c\tau) = D(\omega||\tau) - \log_2 c \tag{10}
\]
for all $c > 0$. For $\rho \in S(\mathcal{H}_{A,B})$, define
\[
\mathcal{R}(\rho) := \inf_{\sigma \in S(\mathcal{H}_{A,B})} (D(\rho||\sigma) + \log_2 \|(\text{id} \otimes T)(\sigma)\|_1)
\]
as the $\mathbb{D}$-Rains relative entropy. Then
\[
\mathcal{R}(\rho) = \inf_{\sigma \in \text{PPT}(\mathcal{H}_{A,B})} D(\rho||\sigma) \tag{11}
\]
and
\[
\mathcal{R}(\rho) = \inf_{\sigma \in \mathcal{L}_+^+(\mathcal{H}_{A,B})} (D(\rho||\sigma) + \log_2 \|(\text{id} \otimes T)(\sigma)\|_1). \tag{12}
\]

**Remark 4** (General Rains relative entropies). Since the property in (10) holds for the quantum relative entropy and for the sandwiched Rényi relative entropy for $\alpha \in (0, 1) \cup (1, \infty)$, the equalities in (11) and (12) hold for the Rains relative entropies constructed from them (as given in (6) and (7)). The same is true for the Rains relative entropies constructed from the Petz and geometric Rényi relative entropies.

**C. Completely PPT-preserving bipartite channels**

We now turn to describing meaningful classes of operations in the bipartite setting. From the outset, we generalize the physically meaningful LOCC, but it should be clear that all that is said applies to this practically important scenario. Let $\mathcal{P} : L(\mathcal{H}_{A,B}) \rightarrow L(\mathcal{H}_{A',B'})$ denote a bipartite quantum channel, with input systems $A$ and $B$ and output systems $A'$ and $B'$. Physically, we think of one party Alice possessing the input system $A$ and the output system $A'$ and another party Bob in a distant laboratory possessing the input system $B$ and the output system $B'$. A bipartite channel $\mathcal{P}$ is completely PPT preserving [39], [40] if the map $(\text{id} \otimes T) \circ \mathcal{P} \circ (\text{id} \otimes T)$ is completely positive. It is well known that
\[
\sigma \in \text{PPT}(\mathcal{H}_{A,B}) \Rightarrow \mathcal{P}(\sigma) \in \text{PPT}(\mathcal{H}_{A',B'}), \tag{13}
\]
and this implies from (3) that
\[
R(\rho) \geq R(\mathcal{P}(\rho)), \quad \tilde{R}_\alpha(\rho) \geq \tilde{R}_\alpha(\mathcal{P}(\rho)), \tag{14}
\]
where the latter holds for all $\alpha \in [1/2, 1) \cup (1, \infty)$. Indeed, to see (13), let $\sigma \in \text{PPT}(\mathcal{H}_{A,B})$ and consider that
\[
\|(\text{id} \otimes T)(\mathcal{P}(\sigma))\|_1 = \|(\text{id} \otimes T) \circ \mathcal{P} \circ (\text{id} \otimes T)((\text{id} \otimes T)(\sigma))\|_1 \leq \|(\text{id} \otimes T)(\sigma)\|_1 \leq 1,
\]
where the first inequality follows because the trace norm does not increase under the action of a completely positive trace-preserving map (in this case $(\text{id} \otimes T) \circ \mathcal{P} \circ (\text{id} \otimes T)$).

The significance of the set of completely PPT-preserving channels [35], [44], [40] is that it contains the operationally relevant set of local operations and classical communication (LOCC) channels [6], [41]. The constraints specifying the former set of channels are semi-definite, and it is thus much easier in a computational sense to optimize an objective function over this set of channels than to optimize over the set of LOCC channels (see, e.g., [24]). Generalizing completely PPT-preserving channels, physically well motivated, and of relevance for us here, is the set of selective PPT operations.

**Definition 5** (Selective PPT operation). The set $\{\mathcal{P}_x\}_x$ constitutes a selective PPT operation if $\mathcal{P}_x \in \text{CP}(\mathcal{H}_{A,B}, \mathcal{H}_{A',B'})$ for all $x$, $(\text{id} \otimes T) \circ \mathcal{P}_x \circ (\text{id} \otimes T) \in \text{CP}(\mathcal{H}_{A,B}, \mathcal{H}_{A',B'}')$ for all $x$, and the sum map $\sum_x \mathcal{P}_x$ is trace preserving.

When performed on a quantum state $\rho \in S(\mathcal{H}_{A,B})$, the selective PPT operation $\{\mathcal{P}_x\}_x$ can be understood as a particular kind of quantum instrument, which outputs the state $\mathcal{P}_x(\rho)/p(x)$ with probability $p(x) = \text{Tr}[\mathcal{P}_x(\rho)]$. Similar to the channel case discussed above, the set of selective PPT operations contains the set of selective LOCC operations [41]. Let $\sigma \in S(\mathcal{H}_{A,B})$ be a PPT state, i.e., a state satisfying $(\text{id} \otimes T)(\sigma) \geq 0$. Let $\sigma_x := \mathcal{P}_x(\sigma)/q(x)$ with $q(x) := \text{Tr}[\mathcal{P}_x(\sigma)]$. It follows that $\sigma_x$ is also a PPT state because
\[
(\text{id} \otimes T)(\sigma_x) = \frac{1}{q(x)} (\text{id} \otimes T) \mathcal{P}_x(\sigma) = \frac{1}{q(x)} (\text{id} \otimes T) \mathcal{P}_x (\text{id} \otimes T)(\sigma) \geq 0,
\]
where the inequality follows because $(\text{id} \otimes T) \circ \mathcal{P}_x (\text{id} \otimes T)$ is completely positive by definition and $(\text{id} \otimes T)(\sigma) \geq 0$. Now suppose that $\sigma \in \text{PPT}(\mathcal{H}_{A,B})$. With $\sigma_x$ and $q(x)$ as defined above, we know that the following inequality holds
\[
\|(\text{id} \otimes T)(\sigma_x)\|_1 \geq \sum_x q(x) \|(\text{id} \otimes T)(\sigma_x)\|_1, \tag{17}
\]
as a consequence of ref. [43, Eq. (8)], following the approach given in ref. [44, Proposition 2.1] (see also ref. [24, Proposition 9.10]). However, it is not clear if $\sigma_x \in \text{PPT}(\mathcal{H}_{A,B})$ for all $x$. This is the main obstacle to overcome in proving Theorem 6 of the next section, and we do so by exploiting Lemma 6 and properties of generalized relative entropies.

**III. SELECTIVE PPT MONOTONICITY**

We now derive one of the main results of this work: All the Rains relative entropies discussed thus far are selective PPT monotones. By the fact that selective LOCC operations are contained in the set of selective PPT operations, it follows
that these quantities are also selective LOCC monotones. Again, as such, they are quantities meaningfully quantifying entanglement, according to the stronger selective notion, in a computable and conservative way. This property is also one of the main ones needed to establish the Rains relative entropies as upper bounds on probabilistic approximate distillable entanglement, which we do in Section IV. We provide a general statement in Theorem 6 below, which applies to the standard, as well as other, Rains relative entropies.

**Theorem 6 (Selective entanglement monotonicity).** Let \( D(\omega \| \tau) \) be a real-valued function of \( \omega \in S(H) \) and \( \tau \in L_+(H) \) for which the data-processing inequality holds:
\[
D(\omega \| \tau) \geq D(\mathcal{N}(\omega) \| \mathcal{N}(\tau))
\]
for \( \mathcal{N} \in \text{CPTP}(H, H') \), and for which the following holds
\[
D(\kappa \| \lambda) \geq D(p \| q) + \sum_x p(x) D(\kappa_x \| \lambda_x),
\]
(18)
for classical–quantum \( \kappa \in S(H_{X,A}) \) and \( \lambda \in L_+(H_{X,A}) \), as defined in Lemma 7. For \( \rho \in S(H_{A|B}) \), define
\[
R(\rho) := \inf_{\sigma \in S(H_{A|B})} (D(\rho \| \sigma) + \log_2(\| \text{id} \otimes T(\sigma) \|_1)),
\]
(19)
as the \( D \)-Rains relative entropy. Then the \( D \)-Rains relative entropy \( R(\rho) \) is a selective PPT monotone; i.e., it satisfies
\[
R(\rho) \geq \sum_{x : p(x) > 0} p(x) R(\rho_x),
\]
where
\[
p(x) := \text{Tr}[P_x(\rho)], \quad \rho_x := \frac{P_x(\rho)}{p(x)},
\]
and \( \{P_x\}_x \) is a selective PPT operation (see Definition 5).

**Proof.** Consider the quantum channel \( P(\cdot) := \sum_x |x\rangle \langle x| \otimes P_x(\cdot) \).

Let \( \sigma \in S(H_{A|B}) \), and let us define the probability distribution \( \{q(x)\}_x \) and set \( \{\sigma_x\}_x \) with \( \sigma_x \in S(H_{A'\otimes B'} \) as
\[
q(x) := \text{Tr}[P_x(\sigma)], \quad \sigma_x := \frac{1}{q(x)} P_x(\sigma).
\]
From the data-processing inequality for \( D \) and (19), we find
\[
D(\rho \| \sigma) \geq D(P(\rho) \| P(\sigma))
\]
\[
= D\left( \sum_x |x\rangle \langle x| \otimes \sum_x |x\rangle \langle x| \otimes P_x(\sigma) \right)
\]
\[
= D\left( \sum_x p(x) |x\rangle \langle x| \otimes \rho_x \right) \sum_x q(x) |x\rangle \langle x| \otimes \sigma_x
\]
\[
\geq D(p \| q) + \sum_x p(x) D(\rho_x \| \sigma_x),
\]
(20)
Putting this together with (17) and defining \( X^+ := \{x : p(x) > 0\} \), we find that
\[
D(\rho \| \sigma) + \log_2(\| \text{id} \otimes T(\sigma) \|_1) \geq \sum_{x \in X^+} p(x) D(\rho_x \| \sigma_x)
\]
\[
+ D(p \| q) + \log_2 \left( \sum_x q(x) \| (\text{id} \otimes T(\sigma)) \|_1 \right)
\]
\[
\geq \sum_{x \in X^+} p(x) D(\rho_x \| \sigma_x) + \sum_{x \in X^+} p(x) \log_2(\| (\text{id} \otimes T(\sigma)) \|_1)
\]
\[
\geq \sum_{x \in X^+} p(x) R(\rho_x).
\]
The inequality follows from the definition in (19). The second inequality follows because
\[
D(p \| q) + \log_2 \left( \sum_x q(x) \| (\text{id} \otimes T(\sigma)) \|_1 \right)
\]
\[
\geq D(p \| q) + \log_2 \left( \sum_{x \in X^+} p(x) q(x) \| (\text{id} \otimes T(\sigma)) \|_1 \right)
\]
\[
geq \sum_{x \in X^+} p(x) \left( \log_2(\frac{q(x)}{p(x)}) + \log_2(\frac{q(x)}{p(x)} \| (\text{id} \otimes T(\sigma)) \|_1) \right)
\]
\[
= \sum_{x \in X^+} p(x) \log_2(\| (\text{id} \otimes T(\sigma)) \|_1),
\]
(22)
where we have used the concavity of the logarithm. Since
\[
D(\rho \| \sigma) + \log_2(\| \text{id} \otimes T(\sigma) \|_1) \geq \sum_{x \in X^+} p(x) R(\rho_x)
\]
holds for arbitrary \( \sigma \in S(H_{A|B}) \), we conclude by taking an infimum over every \( \sigma \in S(H_{A|B}) \) and applying Lemma 5.

**Corollary 7** (Rains relative entropies as selective entanglement monotones). The Rains relative entropies constructed from the quantum relative entropy, and the Petz (for \( \alpha \in (1, 2] \)), geometric (for \( \alpha = (1, 2] \)), and sandwiched (for \( \alpha > 1 \)) Rényi relative entropies are selective PPT monotones.

**Proof.** All quantities satisfy the hypotheses of Theorem 6.

The proof of the following corollary, presented in Appendix C, justifies a claim made in ref. [45] Lemma 24.

**Corollary 8** (Flags property). The Rains relative entropy possesses the “flags” property [2]. That is, for a probability distribution \( \{p(x)\}_x \) and \( \omega_x \in S(H_{A|B}) \), let \( \omega \in S(H_{X_{AB}}) \) be of the form
\[
\omega = \sum_x p(x) |x\rangle \langle x| \otimes \omega_x.
\]
Then, for either of the bipartite cuts \( X|A \otimes B \) or \( A|X \otimes B \),
\[
R(\omega) = \sum_x p(x) R(\omega_x).
\]

**IV. PROBABILISTIC APPROXIMATE DISTILLATION**

In this final section, we turn our attention to an important application of Theorem 6. We prove that the Rains relative entropies are upper bounds on the non-asymptotic and asymptotic probabilistic approximate distillable entanglement (PADE). This result strengthens the original result of ref. [11].
because the PADE cannot be smaller than the standard distillable entanglement [2, 24]. The notion of PADE and such concepts in other resource theories has become increasingly relevant [46, 47, 48, 49, 50, 51], due to the connection with experiment and the fact that desired resource states are in practice generated only approximately and with a certain probability heralded by a classical signal.

A. Non-asymptotic case

Given a bipartite state $\rho \in S(\mathcal{H}_{AB})$ and error parameter $\varepsilon \in [0,1]$, we define the non-asymptotic probabilistic approximate distillable entanglement (PADE) as

$$E^\varepsilon_d(\rho) := \sup_{p \in [0,1], L} \left\{ p \log_2 d : \rho \in L(\rho) = p|1 \rangle |1 \rangle \otimes \tilde{\Phi}^d + (1-p) |0 \rangle |0 \rangle \otimes \sigma , \quad F(\tilde{\Phi}^d, \Phi^d) \geq 1 - \varepsilon , \quad \sigma \in S(\mathcal{H}_{AB}) \right\} ,$$

where the optimization is over $L \in \text{LOCC}(\mathcal{H}_{AB}, \mathcal{H}_{XAB})$, the probability $p \in [0,1]$, and the dimension $d = \dim(\mathcal{H}_{A}) = \dim(\mathcal{H}_{B})$. Also, $F(\omega, \tau) := \|\sqrt{\omega} \sqrt{\tau} \|^2$ is the fidelity of states $\omega, \tau \in S(\mathcal{H})$ [52], and $\tilde{\Phi}^d := [\tilde{\Phi}^d | [\tilde{\Phi}^d]$ is the maximally entangled state,

$$|\tilde{\Phi}^d := d^{-1/2} \sum_i |i\rangle |i\rangle,$$

with $\{|i\rangle\}$, an orthonormal basis. The interpretation of $E^\varepsilon_d(\rho)$ is that it is equal to the expected number of $\varepsilon$-approximate e-bits that can be generated from the state $\rho$ by means of LOCC, where an e-bit is $\tilde{\Phi}^d$. The standard non-asymptotic distillable entanglement of $\rho$ is defined in the same way as above, however, with the exception that there is no optimization over $p$, and it is instead simply set to $p = 1$ [53].

Theorem 9 (Non-asymptotic upper bound). For $\varepsilon \in (0,1)$ and $\rho \in S(\mathcal{H}_{AB})$, the following upper bound holds for all $\alpha > 1$:

$$E^\varepsilon_d(\rho) \leq \tilde{R}_\alpha(\rho) + \frac{\alpha}{\alpha - 1} \log_2 \left( \frac{1}{1 - \varepsilon} \right).$$

Proof. Let $L \in \text{LOCC}(\mathcal{H}_{AB}, \mathcal{H}_{XAB})$ and $p \in [0,1]$ be the elements of an arbitrary non-asymptotic PADE protocol. For $\alpha > 1$, we find that

$$\tilde{R}_\alpha(\rho) \geq p \tilde{R}_\alpha(\tilde{\Phi}^d) + (1-p) \tilde{R}_\alpha(\sigma) \geq p \tilde{R}_\alpha(\tilde{\Phi}^d) \geq p \left[ \frac{\alpha}{\alpha - 1} \log_2 (1 - \varepsilon) + \log_2 d \right] \geq \frac{\alpha}{\alpha - 1} \log_2 (1 - \varepsilon) + p \log_2 d ,$$

where the first inequality follows from applying Theorem 6 and Corollary 7 to $L$ and $\rho$. The second inequality follows because $\tilde{R}_\alpha(\sigma) \geq 0$, and the third inequality follows from applying Lemma 13 in Appendix D. Rewriting the inequality $\tilde{R}_\alpha(\rho) \geq \frac{\alpha}{\alpha - 1} \log_2 (1 - \varepsilon) + p \log_2 d$, we find, for all $\alpha > 1$,

$$p \log_2 d \leq \tilde{R}_\alpha(\rho) + \frac{\alpha}{\alpha - 1} \log_2 \left( \frac{1}{1 - \varepsilon} \right).$$

Since the upper bound depends only on $\rho$, $\varepsilon$, and $\alpha$ and it holds for all $L \in \text{LOCC}(\mathcal{H}_{AB}, \mathcal{H}_{XAB})$ and $p \in [0,1]$, we conclude (25) after applying the definition of $E_d^\varepsilon(\rho)$.

B. Asymptotic case

We define the asymptotic PADE as

$$E_d(\rho) := \inf_{\varepsilon \in (0,1)} \lim_{n \to \infty} \frac{1}{n} E^\varepsilon_d(\rho^\otimes n) ,$$

as well as the strong converse PADE as

$$\tilde{E}_d(\rho) := \sup_{\varepsilon \in (0,1)} \lim_{n \to \infty} \frac{1}{n} E^\varepsilon_d(\rho^\otimes n) .$$

The inequality $E_d(\rho) \leq \tilde{E}_d(\rho)$ is obvious from the definitions.

Theorem 10 (Strong converse). For $\rho \in S(\mathcal{H}_{AB})$, one finds $\tilde{E}_d(\rho) \leq R(\rho)$.

Proof. By applying Theorem 9 we find that the following upper bound holds for all $\varepsilon \in (0,1)$ and $\alpha > 1$.

$$\frac{1}{n} E^\varepsilon_d(\rho^\otimes n) \leq \frac{1}{n} \tilde{R}_\alpha(\rho^\otimes n) + \frac{\alpha}{n(\alpha - 1)} \log_2 \left( \frac{1}{1 - \varepsilon} \right) \leq \frac{1}{n} \tilde{R}_\alpha(\rho) + \frac{\alpha}{n(\alpha - 1)} \log_2 \left( \frac{1}{1 - \varepsilon} \right) ,$$

where we have used subadditivity of the sandwiched Rains relative entropy [9]. Taking the limit as $n \to \infty$, we find that the following holds for all $\varepsilon \in (0,1)$ and $\alpha > 1$

$$\lim_{n \to \infty} \frac{1}{n} E^\varepsilon_d(\rho^\otimes n) \leq \tilde{R}_\alpha(\rho) .$$

The limit $\alpha \to 1$ yields

$$\lim_{n \to \infty} \frac{1}{n} E^\varepsilon_d(\rho^\otimes n) \leq R(\rho) ,$$

where we have used (35). Then, the desired inequality follows because the above bound holds for all $\varepsilon \in (0,1)$.

We can regularize [54] the bound above to make it even tighter, with the proof being presented in Appendix E. In Appendix E, we provide a non-asymptotic bound which leads to a weak-converse upper bound on the PADE.

Corollary 11 (Regularized Rains bound). For $\rho \in S(\mathcal{H}_{AB})$,

$$\tilde{E}_d(\rho) \leq \inf_{\varepsilon \in (0,1)} \frac{1}{n} \tilde{R}(\rho^\otimes \ell) .$$

Remark 12 (Tightest bounds). Due to the subadditivity of the Rains relative entropy [5], the sequence $\{ \frac{1}{n} \tilde{R}(\rho^\otimes \ell) \}_{\ell \in \mathbb{N}}$ is monotone decreasing, so that one obtains upper bounds that become tighter with increasing $\ell$.

In summary, we have proven that various Rains relative entropies are selective PPT entanglement monotones and have established this quantity as a meaningful conservative entanglement measure. We have also shown that they give upper bounds on the probabilistic approximate distillable entanglement, in both the non-asymptotic and asymptotic settings.

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APPENDIX A

PROOF OF LEMMA 2

The inequality in (10) can be understood as a rewriting of ref. [14] Lemma 3]. We give a full proof here for completeness. Consider that
\[
\bar{D}_\alpha(\kappa || \lambda) = \frac{1}{\alpha - 1} \log_2 \left( \sum_x \rho^\alpha(x) q_1 - \alpha(x) \tilde{Q}_\alpha(\kappa_x || \lambda_x) \right)
\]
and by applying (10). Since the inequality holds for all \( \sigma \in S(\mathcal{H}_{AB}) \), we conclude that
\[
\inf_{\sigma \in \text{PPT}'(\mathcal{H}_{AB})} \mathcal{D}(\rho || \sigma) \leq \inf_{\omega \in \mathcal{S}(\mathcal{H}_{AB})} \left[ \mathcal{D}(\rho || \omega) + \log_2 \left( \| \text{id} \otimes T(\omega) \|_1 \right) \right].
\] (36)
The first equality follows from (10) and the first inequality from the fact that \( \| (\text{id} \otimes T)(\sigma) \|_1 \leq 1 \). Since the inequality holds for arbitrary \( \sigma \in \text{PPT}'(\mathcal{H}_{AB}) \), we conclude that
\[
\inf_{\sigma \in \text{PPT}'(\mathcal{H}_{AB})} \mathcal{D}(\rho || \sigma) = \inf_{\omega \in \mathcal{S}(\mathcal{H}_{AB})} \left[ \mathcal{D}(\rho || \omega) + \log_2 \left( \| (\text{id} \otimes T)(\omega) \|_1 \right) \right].
\] (37)
Putting together (35) and (37), we conclude the desired equality in (25). The proof of the equality expressed in (14) is the same.

APPENDIX C

PROOF OF COROLLARY 8

Convexity of \( R \) and its monotonicity under the local channel \( \cdot \mapsto |x\rangle\langle x| \otimes \cdot \) (applying (13)) imply that
\[
R(\omega) \leq \sum_x p(x) R(|x\rangle\langle x| \otimes \omega_x) \leq \sum_x p(x) R(\omega_x).
\]
Defining the local projection \( \tilde{\Sigma}_x : \mathcal{L}(\mathcal{H}_X) \rightarrow \mathcal{L}(\mathcal{H}_X) \) as
\[
\cdot \mapsto |x\rangle\langle x| \cdot |x\rangle\langle x|,
\]
and observing that the sum map \( \sum_x \tilde{\Sigma}_x \) is trace preserving implies that the set \( \{ \tilde{\Sigma}_x \otimes \text{id} : \mathcal{L}(\mathcal{H}_{X AB}) \rightarrow \mathcal{L}(\mathcal{H}_{X AB}) \}_{x} \) constitutes a selective PPT operation. Applying Theorem 6 we find that
\[
R(\omega) \geq \sum_x p(x) R(|x\rangle\langle x| \otimes \omega_x) \geq \sum_x p(x) R(\omega_x),
\]
where the last inequality follows because the partial trace over \( \mathcal{H}_X \) is a local channel and \( R \) is monotone under the action of local channels.

APPENDIX D

PROOF OF INEQUALITY FOR RAINS RELATIVE ENTROPY OF APPROXIMATE E-BITS

In this appendix, we prove Lemma 13 which states that the sandwiched Rains relative entropy and a correction term provide an upper bound on the number of approximate e-bits that are available in a state \( \tilde{\Phi}^d \) close in fidelity to an ideal maximally entangled state \( \Phi^d \). We provide two different proofs of this lemma: a first that is similar to ref. [27] Proposition 4] and a second in terms of a pseudo-continuity bound for the sandwiched Rains relative entropy, given in Lemma 14 below.

**Lemma 13 (Approximate e-bits).** Let \( \varepsilon \in [0, 1] \), \( \tilde{\Phi}^d \in S(\mathcal{H}_{AB}) \), and \( \Phi^d \in S(\mathcal{H}_{AB}) \), the last defined through (24). Suppose that \( F(\tilde{\Phi}^d, \Phi^d) \geq 1 - \varepsilon \). Then, for all \( \alpha > 1 \),
\[
\log_2 d \leq \bar{R}_\alpha(\tilde{\Phi}^d) + \frac{\alpha}{\alpha - 1} \log_2 \left( \frac{1}{1 - \varepsilon} \right).
\]
**Proof.** We provide two different proofs of this statement. The first proof is similar to that given in ref. [37 Proposition 4], and we provide it for completeness. Let a measurement channel \( \mathcal{M} \in \text{CPTP}(\mathcal{H}_{\tilde{A}\tilde{B}}, \mathcal{H}_X) \) be defined as
\[
\mathcal{M}(\cdot) \coloneqq \text{Tr}[\Phi^d(\cdot)]|1\rangle\langle 1| + \text{Tr}[(I - \Phi^d(\cdot))]|0\rangle\langle 0|.
\]
Recall that \( F(\tilde{\Phi}^d, \Phi^d) = \text{Tr}[\Phi^d \tilde{\Phi}^d] \) because \( \Phi^d \) is pure. Then, for \( \sigma \in \text{PPT}'(\mathcal{H}_{\tilde{A}\tilde{B}}) \) and \( \alpha > 1 \), we find that
\[
\tilde{D}_\alpha(\tilde{\Phi}^d||\sigma) \\
\geq \tilde{D}_\alpha(M(\tilde{\Phi}^d)||\mathcal{M}(\sigma)) \\
= \frac{1}{\alpha - 1} \log_2 \left[ \frac{\left( \text{Tr}[\Phi^d \tilde{\Phi}^d] \right)^\alpha}{\left( \text{Tr}[\Phi^d \sigma] - \text{Tr}[\Phi^d \tilde{\Phi}^d] \right)^{1-\alpha}} \right] \\
\geq \frac{1}{\alpha - 1} \log_2 \left[ \frac{\left( \text{Tr}[\Phi^d \tilde{\Phi}^d] \right)^\alpha}{\left( \text{Tr}[\Phi^d \sigma] - \text{Tr}[\Phi^d \tilde{\Phi}^d] \right)^{1-\alpha}} \right] \\
\geq \frac{1}{\alpha - 1} \log_2 \left[ \frac{1}{1 - \varepsilon} \right] \\
= -\frac{\alpha}{\alpha - 1} \log_2 \left( \frac{1}{1 - \varepsilon} \right) + \log_2 d. \tag{38}
\]
The first inequality follows from the data-processing inequality for \( \tilde{D}_\alpha \) for \( \alpha > 1 \). The second inequality follows by dropping the second term inside the logarithm. The third inequality follows by applying the assumption that \( F(\tilde{\Phi}^d, \Phi^d) \geq 1 - \varepsilon \), as well as the known fact that \([55 \text{ Lemma 2}]
\]
\[
\text{Tr}[\Phi^d \sigma] \leq \frac{1}{d} \tag{39}
\]
for \( \sigma \in \text{PPT}'(\mathcal{H}_{\tilde{A}\tilde{B}}) \). Indeed, observing that the set \( \text{PPT}'(\mathcal{H}_{\tilde{A}\tilde{B}}) \) is the same regardless of the basis in which the partial transpose \( \text{id} \otimes T \) is taken, we can take the basis to be the same as that for the maximally entangled state \( \Phi^d \), and we find that
\[
\text{Tr}[\Phi^d \sigma] = \text{Tr}[\Phi^d ((\text{id} \otimes T)(\text{id} \otimes T))\sigma] \\
= \text{Tr}[(\text{id} \otimes T)(\Phi^d)(\text{id} \otimes T)(\sigma)] \\
= \frac{1}{d} \text{Tr}[\Phi^d (\text{id} \otimes T)(\sigma)] \\
\leq \frac{1}{d} ||(\text{id} \otimes T)(\sigma)||_1 \leq \frac{1}{d}. \tag{40}
\]
In the above, we used the facts that the partial transpose is self-inverse, self-adjoint with respect to the Hilbert–Schmidt inner product, that \((\text{id} \otimes T)(\Phi^d) = \frac{1}{d} F^d\), where \( F^d \in \mathcal{L}(\mathcal{H}_{\tilde{A}\tilde{B}}) \) is the unitary swap operator, and the variational characterization of the trace norm of \( X \in \mathcal{L}(\mathcal{H}) \) as \( \|X\|_1 = \sup_{U \in \mathcal{L}(\mathcal{H})} \text{Tr}[X U] \), with the supremum over every unitary \( U \in \mathcal{L}(\mathcal{H}) \).

Returning to (38), we have thus proven the following inequality for all \( \sigma \in \text{PPT}'(\mathcal{H}_{\tilde{A}\tilde{B}}) \) and \( \alpha > 1 \):
\[
\tilde{D}_\alpha(\tilde{\Phi}^d||\sigma) \\
\geq -\frac{\alpha}{\alpha - 1} \log_2 \left( \frac{1}{1 - \varepsilon} \right) + \log_2 d.
\]
We conclude after taking an infimum over \( \sigma \in \text{PPT}'(\mathcal{H}_{\tilde{A}\tilde{B}}) \) and applying the definition of \( \tilde{R}_\alpha(\Phi^d) \).

A second proof follows by applying Lemma [14] below:
\[
\tilde{R}_\alpha(\tilde{\Phi}^d) \geq \tilde{R}_\beta(\Phi^d) + \frac{\alpha}{\alpha - 1} \log_2 F(\Phi^d, \Phi^d) \\
\geq \tilde{R}_\beta(\Phi^d) + \frac{\alpha}{\alpha - 1} \log_2 (1 - \varepsilon) \\
\geq \tilde{R}_{1/2}(\Phi^d) + \frac{\alpha}{\alpha - 1} \log_2 (1 - \varepsilon), \tag{41}
\]
where we have employed the monotonicity of the sandwiched Rényi relative entropy with respect to \( \alpha \) [24]. Then consider that
\[
\tilde{R}_{1/2}(\Phi^d) = \inf_{\sigma \in \text{PPT}'(\mathcal{H}_{\tilde{A}\tilde{B}})} \left[ -\log_2 F(\Phi^d, \sigma) \right] \\
= -\log_2 \sup_{\sigma \in \text{PPT}'(\mathcal{H}_{\tilde{A}\tilde{B}})} F(\Phi^d, \sigma) \\
\geq -\log_2 \sup_{\sigma \in \text{PPT}'(\mathcal{H}_{\tilde{A}\tilde{B}})} \text{Tr}[\Phi^d \sigma] \\
\geq \log_2 d. \tag{42}
\]
where we again have applied (39).

**Lemma 14** (Pseudo-continuity bound). For states \( \rho_0, \rho_1 \in S(\mathcal{H}_{\tilde{A}\tilde{B}}), \beta \in (1/2, 1), \) and \( \alpha = \beta/(2\beta - 1) > 1 \), the following bound holds
\[
\tilde{R}_\alpha(\rho_0) - \tilde{R}_\beta(\rho_1) \geq \frac{\alpha}{\alpha - 1} \log_2 F(\rho_0, \rho_1). \tag{43}
\]

**Proof.** Recall the following bound from ref. [55 Lemma 1]:
\[
\tilde{D}_\alpha(\rho_0||\sigma) - \tilde{D}_\beta(\rho_1||\sigma) \geq \frac{\alpha}{\alpha - 1} \log_2 F(\rho_0, \rho_1),
\]
which holds for \( \rho_0, \rho_1 \in S(\mathcal{H}_{\tilde{A}\tilde{B}}), \sigma \in \mathcal{L}_+(\mathcal{H}_{\tilde{A}\tilde{B}}), \beta \in (1/2, 1), \) and \( \alpha = \beta/(2\beta - 1) > 1 \). (An inspection of the proof there reveals that it holds more generally for \( \sigma \in \mathcal{L}_+(\mathcal{H}_{\tilde{A}\tilde{B}}) \), and not just for \( \sigma \in S(\mathcal{H}_{\tilde{A}\tilde{B}}) \).) Then, for \( \sigma \in \text{PPT}'(\mathcal{H}_{\tilde{A}\tilde{B}}) \), we can rewrite this as
\[
\tilde{D}_\alpha(\rho_0||\sigma) \geq \tilde{D}_\beta(\rho_1||\sigma) + \frac{\alpha}{\alpha - 1} \log_2 F(\rho_0, \rho_1) \\
\geq \inf_{\sigma \in \text{PPT}'(\mathcal{H}_{\tilde{A}\tilde{B}})} \tilde{D}_\beta(\rho_1||\sigma) + \frac{\alpha}{\alpha - 1} \log_2 F(\rho_0, \rho_1) \\
= \tilde{R}_\beta(\rho_1) + \frac{\alpha}{\alpha - 1} \log_2 F(\rho_0, \rho_1). \tag{44}
\]
Since this inequality holds for every \( \sigma \in \text{PPT}'(\mathcal{H}_{\tilde{A}\tilde{B}}) \), we conclude that
\[
\inf_{\sigma \in \text{PPT}'(\mathcal{H}_{\tilde{A}\tilde{B}})} \tilde{D}_\alpha(\rho_0||\sigma) \geq \tilde{R}_\beta(\rho_1) + \frac{\alpha}{\alpha - 1} \log_2 F(\rho_0, \rho_1),
\]
which is equivalent to the desired inequality in [43].

**APPENDIX E**

**PROOF OF COROLLARY 11**

We begin with the following lemma.

**Lemma 15** (Dimension bound). For \( \varepsilon \in (0, 1) \) and \( \rho \in S(\mathcal{H}_{\tilde{A}\tilde{B}}) \), the following bounds hold
\[
0 \leq E_\varepsilon^\alpha(\rho) \leq \log_2 D + \log_2 \left( \frac{1}{1 - \varepsilon} \right),
\]
where
\[
D := \min \{ \dim(\mathcal{H}_A), \dim(\mathcal{H}_B) \}. \]
Proof. The lower bound on $E_d^\ell(\rho)$ is a direct consequence of its definition. For the other bound, we employ Theorem 9 to conclude that the following holds for all $\alpha > 1$:

\[ E_d^\alpha(\rho) \leq \tilde{R}_\alpha(\rho) + \frac{\alpha}{\alpha - 1} \log_2 \left( \frac{1}{1 - \varepsilon} \right) \]
\[ \leq \tilde{R}_\alpha(\Phi^D) + \frac{\alpha}{\alpha - 1} \log_2 \left( \frac{1}{1 - \varepsilon} \right) \]
\[ \leq \log_2 D + \frac{\alpha}{\alpha - 1} \log_2 \left( \frac{1}{1 - \varepsilon} \right). \]

(45)

The second inequality follows from a teleportation argument: $\rho$ can be realized from $\Phi^D$ by means of LOCC via the teleportation protocol (see, e.g., ref. [23]). The final inequality is a consequence of the same reasoning given for Eq. (5.2.88) of ref. [24]. Since the bound holds for all $\alpha > 1$, we can take the limit $\alpha \to \infty$ to arrive at the desired upper bound on $E_d^\ell(\rho)$.

We now state the following theorem.

**Theorem 16 (Regularized quantities).** For all $\varepsilon \in (0, 1)$, $\rho \in \mathcal{S}(H_{AB})$, and $\ell \in \mathbb{N}$, the following equalities hold

\[ \liminf_{n \to \infty} \frac{1}{n} E_d^\ell(\rho^{\otimes n}) = \frac{1}{\ell} \liminf_{n \to \infty} \frac{1}{n} E_d^\ell(\rho^{\otimes k \ell}) \]
\[ \limsup_{n \to \infty} \frac{1}{n} E_d^\ell(\rho^{\otimes n}) = \frac{1}{\ell} \limsup_{n \to \infty} \frac{1}{n} E_d^\ell(\rho^{\otimes k \ell}). \]

(46)

(47)

Consequently, for all $\varepsilon \in (0, 1)$,

\[ E_d(\rho) = \frac{1}{\ell} E_d(\rho^{\otimes k \ell}), \]
\[ \tilde{E}_d(\rho) = \frac{1}{\ell} \tilde{E}_d(\rho^{\otimes k \ell}). \]

(48)

(49)

**Proof.** First consider that

\[ \liminf_{n \to \infty} \frac{1}{n} E_d^\ell(\rho^{\otimes n}) \leq \frac{1}{\ell} \liminf_{n \to \infty} \frac{1}{n} E_d^\ell(\rho^{\otimes k \ell}), \]
\[ \limsup_{n \to \infty} \frac{1}{n} E_d^\ell(\rho^{\otimes n}) \geq \frac{1}{\ell} \limsup_{n \to \infty} \frac{1}{n} E_d^\ell(\rho^{\otimes k \ell}), \]

(50)

(51)

because \{1/n \ E_d^\ell(\rho^{\otimes n})\}_n \ \in \mathbb{N} \ is \ a \ subsequence \ of \ \{1/n \ E_d^\ell(\rho^{\otimes n})\}_n \ \in \mathbb{N}.

So it remains to prove the opposite inequalities. We first show that

\[ \limsup_{n \to \infty} \frac{1}{n} E_d^\ell(\rho^{\otimes n}) \leq \frac{1}{\ell} \limsup_{n \to \infty} \frac{1}{n} E_d^\ell(\rho^{\otimes k \ell}). \]

(52)

To this end, we make use of an idea from the proof of Theorem 8 in ref. [37]. For a fixed positive integer $\ell$, consider that

\[ \limsup_{n \to \infty} \frac{1}{n} E_d^\ell(\rho^{\otimes n}) = \max_{j \in \{0, 1, \ldots, \ell - 1\}} \limsup_{k \to \infty} \frac{1}{k \ell + j} E_d^\ell(\rho^{\otimes (k \ell + j)}). \]

(53)

Consider now a fixed $j \in \{0, 1, \ldots, \ell - 1\}$. Then we find that

\[ \frac{1}{k \ell + j} E_d^\ell(\rho^{\otimes (k \ell + j)}) \leq \frac{1}{k \ell + j} E_d^\ell(\rho^{\otimes ((k+1)\ell)}) \]
\[ = \frac{1}{(k+1)\ell} E_d^\ell(\rho^{\otimes ((k+1)\ell)}) \]
\[ + \frac{\ell - j}{(k+1)\ell (k \ell + j)} E_d^\ell(\rho^{\otimes ((k+1)\ell)}). \]

(54)

We conclude that

\[ \limsup_{k \to \infty} \frac{\ell - j}{(k+1)\ell (k \ell + j)} E_d^\ell(\rho^{\otimes ((k+1)\ell)}) = 0 \]
because, from Lemma [18]

\[ 0 \leq \frac{\ell - j}{(k+1)\ell (k \ell + j)} E_d^\ell(\rho^{\otimes ((k+1)\ell)}), \]
and

\[ \frac{\ell - j}{(k+1)\ell (k \ell + j)} E_d^\ell(\rho^{\otimes ((k+1)\ell)}) \]
\[ \leq \frac{\ell - j}{k \ell + j} \left[ \log_2 D + \frac{1}{(k+1)\ell} \log_2 \left( \frac{1}{1 - \varepsilon} \right) \right], \]
so that

\[ \limsup_{k \to \infty} \frac{\ell - j}{k \ell + j} \log_2 D + \frac{1}{(k+1)\ell} \log_2 \left( \frac{1}{1 - \varepsilon} \right) = 0. \]

Thus, we find that

\[ \limsup_{k \to \infty} \frac{1}{k \ell + j} E_d^\ell(\rho^{\otimes (k \ell + j)}) \]
\[ \leq \limsup_{k \to \infty} \frac{1}{k \ell + j} E_d^\ell(\rho^{\otimes ((k+1)\ell)}) \]
\[ = \frac{1}{k+1} \limsup_{k \to \infty} \frac{1}{k+1} E_d^\ell(\rho^{\otimes ((k+1)\ell)}). \]

(55)

Since this upper bound is independent of $j$, we combine with (53) to arrive at the desired inequality in (52). The inequality

\[ \liminf_{n \to \infty} \frac{1}{n} E_d^\ell(\rho^{\otimes n}) \geq \frac{1}{\ell} \liminf_{n \to \infty} \frac{1}{n} E_d^\ell(\rho^{\otimes n}) \]
can be proved in a similar fashion, by using the fact that

\[ \liminf_{n \to \infty} \frac{1}{n} E_d^\ell(\rho^{\otimes n}) = \min_{j \in \{0, 1, \ldots, \ell - 1\}} \liminf_{k \to \infty} \frac{1}{k \ell + j} E_d^\ell(\rho^{\otimes (k \ell + j)}). \]

(56)

Thus we conclude the desired equalities in (46)–(47). Since the equalities in (46)–(47) hold for all $\varepsilon \in (0, 1)$, we conclude (48)–(49) after applying the definitions in (27)–(28), respectively.

**Proof of Corollary 17** Apply (49) and then Theorem 11 to the state $\rho^{\otimes \ell}$ to conclude that

\[ \tilde{E}_d(\rho) = \frac{1}{\ell} \tilde{E}_d(\rho^{\otimes \ell}) \leq \frac{1}{\ell} R(\rho^{\otimes \ell}). \]

Since this upper bound holds for every $\ell \in \mathbb{N}$, the infimum over $\ell \in \mathbb{N}$ is an upper bound as well.
APPENDIX F

WEAK CONVERSE BOUND ON PADE

In this appendix, we establish a bound on the non-asymptotic PADE, which can be used to arrive at a weak converse bound on the asymptotic PADE.

A. Upper bound on non-asymptotic PADE

**Theorem 17** (Single-shot weak converse). For \( \varepsilon \in [0, 1/2] \) and \( \rho \in S(H_{AB}) \), the following upper bound holds,

\[
E_d^*(\rho) \leq \frac{1}{1 - \varepsilon} \left[ R(\rho) + h_2(\varepsilon) \right]. \tag{57}
\]

**Proof.** Let \( L \in \text{LOCC}(H_{AB}, H_X \otimes \tilde{H}_B) \) and \( \rho \in [0, 1] \) be the elements of an arbitrary non-asymptotic PADE protocol. Consider that

\[
R(\rho) \geq pR(\tilde{\Phi}^d) + (1 - p)R(\sigma) \geq pR(\tilde{\Phi}^d) \geq p \left[ (1 - \varepsilon) \log_2 d - h_2(\varepsilon) \right] \geq p \left( 1 - \varepsilon \right) \log_2 d - h_2(\varepsilon), \tag{58}
\]

where the first inequality follows from applying Theorem 6 and Corollary 7 to \( L \) and \( \rho \). The second inequality follows because \( R(\sigma) \geq 0 \), and the third inequality follows from applying Lemma 18 below. Rewriting the inequality \( R(\rho) \geq p(1 - \varepsilon) \log_2 d - h_2(\varepsilon) \), we find that

\[
p \log_2 d \leq \frac{1}{1 - \varepsilon} \left[ R(\rho) + h_2(\varepsilon) \right].
\]

Since the upper bound depends only on \( \rho \) and \( \varepsilon \) and it holds for all \( L \in \text{LOCC}(H_{AB}, H_X \otimes \tilde{H}_B) \) and \( \rho \in [0, 1] \), we conclude (57) after applying the definition of \( E_d^*(\rho) \).

**Lemma 18** (Weak convexity). Let \( \varepsilon \in [0, 1/2] \), \( \tilde{\Phi}^d \in S(H_{\tilde{A}B}) \), and \( \Phi^d \in S(H_{\tilde{A}B}) \) as defined in (24). Suppose that \( F(\tilde{\Phi}^d, \Phi^d) \geq 1 - \varepsilon \). Then

\[
(1 - \varepsilon) \log_2 d \leq R(\tilde{\Phi}^d) + h_2(\varepsilon).
\]

**Proof.** Let a measurement channel \( \mathcal{M} \in \text{CPTP}(H_{\tilde{A}B}, H_X) \) be defined as

\[
\mathcal{M}(\cdot) := \text{Tr}[\Phi^d(\cdot)]|1\rangle|1\rangle + \text{Tr}[(I - \Phi^d)(\cdot)]|0\rangle|0\rangle.
\]

Recall that \( F(\tilde{\Phi}^d, \Phi^d) = \text{Tr}[\tilde{\Phi}^d \Phi^d] \) because \( \Phi^d \) is pure. Then, for \( \sigma \in \text{PPT}'(H_{\tilde{A}B}) \), we find that

\[
D(\tilde{\Phi}^d||\sigma) \geq D(M(\tilde{\Phi}^d)||M(\sigma)) = \text{Tr}[\Phi^d \tilde{\Phi}^d] \log_2 \left( \frac{\text{Tr}[\Phi^d \tilde{\Phi}^d]}{\text{Tr}[\Phi^d \sigma]} \right) + (1 - \text{Tr}[\Phi^d \tilde{\Phi}^d]) \log_2 \left( \frac{1 - \text{Tr}[\Phi^d \tilde{\Phi}^d]}{\text{Tr}[\sigma - \text{Tr}[\Phi^d \sigma]]} \right)
\]

\[
= -h_2(1 - \text{Tr}[\Phi^d \tilde{\Phi}^d]) - \text{Tr}[\Phi^d \tilde{\Phi}^d] \log_2 \text{Tr}[\Phi^d \sigma] - (1 - \text{Tr}[\Phi^d \tilde{\Phi}^d]) \log_2 (\text{Tr}[\sigma - \text{Tr}[\Phi^d \sigma]])
\]

\[
\geq -h_2(\varepsilon) + (1 - \varepsilon) \log_2 d. \tag{59}
\]

The first inequality follows from the data-processing inequality for \( D \). The second inequality follows because

\[
- \left(1 - \text{Tr}[\Phi^d \tilde{\Phi}^d]\right) \log_2 (\text{Tr}[\sigma] - \text{Tr}[\Phi^d \sigma]) \geq 0.
\]

Also, we have applied the assumption that \( 1 - \text{Tr}[\Phi^d \tilde{\Phi}^d] \leq \varepsilon \) and that the binary entropy function is monotone increasing on the interval \([0, 1/2]\). Additionally, we applied (59) Since the inequality \( D(\tilde{\Phi}^d||\sigma) \geq -h_2(\varepsilon) + (1 - \varepsilon) \log_2 d \) holds for all \( \sigma \in \text{PPT}'(H_{\tilde{A}B}) \), we conclude the desired inequality after taking the infimum over \( \sigma \in \text{PPT}'(H_{\tilde{A}B}) \).

B. Weak-converse upper bound on asymptotic PADE

Here we show that the asymptotic PADE is bounded from above by the Rains relative entropy as

\[
E_d(\rho) \leq R(\rho).
\]

This inequality actually follows directly from Theorem 10 and the fact that \( E_d(\rho) \leq \bar{E}_d(\rho) \), but here we see it as a direct consequence of Theorem 17. Indeed, consider that

\[
E_d(\rho) = \inf_{\varepsilon \in (0, 1)} \liminf_{n \to \infty} \frac{1}{n} E_d^*(\rho^{\otimes n}) \leq \inf_{\varepsilon \in (0, 1)} \liminf_{n \to \infty} \left( \frac{1}{n(1 - \varepsilon)} \left[ R(\rho^\otimes n) + h_2(\varepsilon) \right] \right) \leq \inf_{\varepsilon \in (0, 1)} \liminf_{n \to \infty} \left( \frac{1}{1 - \varepsilon} \left[ R(\rho) + h_2(\varepsilon) \right] \right) = \inf_{\varepsilon \in (0, 1)} \frac{1}{1 - \varepsilon} R(\rho) = R(\rho). \tag{60}
\]

For the first inequality, we have applied Theorem 17 and for the second, we have made use of the subadditivity of the Rains relative entropy. Employing the same kind of argument from Appendix E, we arrive at

\[
E_d(\rho) \leq \inf_{\varepsilon \in \mathbb{N}} R(\rho^\otimes \varepsilon)
\]

as a regularized Rains relative entropy bound.