# Second order asymptotics of visible mixed quantum source coding via universal codes* 

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#### Abstract

The simplest example of a quantum information source with memory is a mixed source which emits signals entirely from one of two memoryless quantum sources with given a priori probabilities. Considering a mixed source consisting of a general one-parameter family of memoryless sources, we derive the second order asymptotic rate for fixed-length visible source coding. Furthermore, we specialize our main result to a mixed source consisting of two memoryless sources. Our results provide the first example of second order asymptotics for a quantum information-processing task employing a resource with memory. For the case of a classical mixed source (using a finite alphabet), our results reduce to those obtained by Nomura and Han [16]. To prove the achievability part of our main result, we introduce universal quantum source codes achieving second order asymptotic rates. These are obtained by an extension of Hayashi's construction [11] of their classical counterparts.


## 1 Introduction

Source coding (or data compression) is essential for efficient storage and transmission of information. Hence, evaluating the optimal rate of data compression is a fundamental problem in information theory. In classical information theory, the simplest class of sources is composed of so-called i.i.d. or stationary, memoryless sources, the name 'memoryless' arising from the fact that there is no correlation between successive signals emitted by such a source. Although these sources play a prominent role in information theory, in real-world applications the assumption of sources being memoryless is not necessarily justified. This is why it is important to study data compression for sources with memory. The simplest example of such a source is a mixed source. It can be constructed from two i.i.d. sources as follows. One associates a priori probabilities, say $t$ and $(1-t)$, to the two sources respectively. Then the mixed source is one for which all successive signals are emitted from the first source with probability $t$, or from the second source with probability $(1-t)$. The memory of the mixed source can be trivially seen to be governed by a two-state Markov chain which is aperiodic but not irreducible, and hence such a source is non-ergodic (see e.g. [17]).

Optimal rates of reliable data compression for the above sources and their quantum analogues were originally evaluated under the requirement that the error incurred in the compression

[^0]and decompression scheme vanishes in the asymptotic limit (i.e. the limit $n \rightarrow \infty$ where $n$ denotes the number of uses of the source). The optimal asymptotic rate for a classical i.i.d. source is given by its Shannon entropy [21], whereas the corresponding rate for a quantum memoryless source is given by its von Neumann entropy [20]. The optimal (first order) asymptotic rate for mixed source coding was derived by Han [9] in the classical case, and in [4] in the quantum case, employing the so-called Information Spectrum Approach. ${ }^{1}$ It was shown to be given by the maximum of the Shannon (resp. von Neumann) entropies of the two underlying classical (resp. quantum) memoryless sources.

Recently, a more refined asymptotic analysis of data compression for memoryless sources under the (more reasonable) requirement of a non-zero error threshold $\varepsilon \in(0,1)$ was done ([7], see also [24]). The quantity analysed was the minimum compression length, which we denote by $\log M_{n} \equiv \log _{2} M_{n}$. In the classical case this is the minimum number of bits needed to compress signals emitted by $n$ uses of the source so that they can be recovered with an error of at most $\varepsilon$ upon decompression. In the quantum case, it is the minimum dimension of the compressed Hilbert space compatible with the given error threshold. The second order asymptotic expansions of the minimum compression length for both the classical and quantum cases were proved to be of the form

$$
\begin{equation*}
\log M_{n}=a n+b \sqrt{n}+\mathcal{O}(\log n) . \tag{1.1}
\end{equation*}
$$

Here, the coefficient $a$ of the leading order term constitutes the first order asymptotics of the minimum compression length, and, as expected, is given by the optimal asymptotic rate. The coefficient $b$ is a function of both the source and the allowed error threshold $\varepsilon$. It constitutes the second order asymptotics and is hence referred to as the second order asymptotic rate (cf. Definition 3.2). It is given by $-\sqrt{V} \Phi^{-1}(\varepsilon)$, where $\Phi^{-1}$ denotes the inverse of the cumulative distribution function of the standard normal distribution (defined in (2.1)), and $V$ denotes the information variance of the source (cf. Definition 2.2(ii)). The asymptotic expansion (1.1) was evaluated for fixed-length source coding in the classical case by Strassen [22] (see also Hayashi [11]) and in the quantum case (for the visible setting) in [7].

Deriving second order asymptotic rates in Classical Information Theory was initiated by Strassen [22]. In Quantum Information Theory, the topic was introduced in 2012 independently by Li [15] and Tomamichel and Hayashi [24], who obtained a second order asymptotic characterization of hypothesis testing. In the latter paper, the authors used this result to characterize the second order asymptotics of randomness extraction and source compression with quantum side information. Since then second order asymptotic expansions have been obtained for a range of operational quantities characterizing information-processing tasks. These include entanglement conversion $[13,7]$, classical-quantum channel coding [25, 3, 7], quantum source coding [7], source coding with quantum side information [24, 3], noisy dense-coding [7], achievability bounds on the coding rate for entanglement-assisted communication [8], an achievability bound on the quantum communication cost in state redistribution [6], and achievability bounds on the quantum capacity [ 2,23 ]. Common to all these endeavours is that the underlying resource (such as the source state in source coding, or the channel in classical-quantum channel coding) is assumed to be memoryless.

Obtaining second order asymptotic expansions for any information-processing task employing resources with memory is a more challenging task. The first foray into this task was made in classical information theory by Polyanskiy, Poor and Verdú [19], who obtained second order expansions for the capacity of a classical mixed channel (see also [26]). In [16], Nomura and

[^1]Han evaluated second order optimal rates for fixed-length source coding for a classical mixed source (see also [11]). Yagi and Nomura [31] (see also Yagi, Han, Nomura [30]) derived the second order coding rate, or channel dispersion, of a mixed channel under the assumption that the channel is well-ordered (cf. [31, Def. 3] or [30, Def. 3]).

All the works mentioned above emphasize the importance of mixed source coding or mixed channel coding as simple yet instructive examples of an information-theoretic task employing non-ergodic resources. The main focus of our paper is to extend the analysis of such tasks to the quantum regime, by investigating mixed quantum source coding. We consider fixed-length source coding for a mixed source constructed from a general one-parameter family of memoryless sources, obtaining optimal second order rates in the visible setting. In the classical case, our results reproduce the optimal rates of Nomura and Han in the finite-alphabet setting. The key tool in our derivations is the second order asymptotic expansion of the information spectrum entropy $D_{s}^{\varepsilon}(\rho \| \tau)$ (see (2.3) for a definition), which was derived in [24]. To prove achievability of the second order asymptotic rates, we introduce universal quantum source codes achieving second order asymptotic rates. These universal codes are obtained by extending the original construction of universal quantum source codes by Jozsa et al. [12] using Hayashi's construction of classical universal source codes which achieve second order asymptotic rates [11].

The paper is organized as follows. After setting the notation and providing the necessary mathematical prerequisites in Section 2, we discuss the operational setting of mixed source coding in Section 3: In Section 3.1 we explain in detail how a mixed source consisting of a one-parameter family of memoryless sources is constructed. Section 3.2 gives a short overview of visible quantum source coding. In Section 3.3 we define the second order asymptotic rate of a quantum source. Our main result is given in Section 4 and comprises expressions for the second order asymptotic rates of mixed source coding. The proofs of these expressions are given in Section 5. For the achievability proofs, we construct universal source codes achieving second order rates in Section 5.1. Finally, in Section 6 we present a conclusion and mention open problems.

## 2 Mathematical preliminaries

For a Hilbert space $\mathcal{H}$, let $\mathcal{B}(\mathcal{H})$ denote the algebra of linear operators acting on $\mathcal{H}$, and let $\mathcal{P}(\mathcal{H})$ denote the set of positive semi-definite operators on $\mathcal{H}$. Further, let $\mathcal{D}(\mathcal{H}):=\{\rho \in \mathcal{P}(\mathcal{H}) \mid$ $\operatorname{Tr} \rho=1\}$ denote the set of states (density matrices) on $\mathcal{H}$. For a state $\rho \in \mathcal{D}(\mathcal{H})$, the von Neumann entropy $S(\rho)$ is defined as $S(\rho):=-\operatorname{Tr}(\rho \log \rho)$. Here and henceforth, all logarithms are taken to base 2, and all Hilbert spaces are assumed to be finite-dimensional. We denote by $\mathbb{1} \in \mathcal{P}(\mathcal{H})$ the identity operator on $\mathcal{H}$, and by id: $\mathcal{B}(\mathcal{H}) \rightarrow \mathcal{B}(\mathcal{H})$ the identity map on operators on $\mathcal{H}$. For a pure state $|\psi\rangle$, the corresponding projector is abbreviated as $\psi \equiv|\psi\rangle\langle\psi|$.

A quantum operation $\Lambda: \mathcal{D}(\mathcal{H}) \rightarrow \mathcal{D}\left(\mathcal{H}^{\prime}\right)$ is a linear, completely positive, trace-preserving (CPTP) map. For self-adjoint operators $A, B \in \mathcal{B}(\mathcal{H})$, let $\{A \geq B\}$ denote the projector onto the subspace spanned by the eigenvectors of the operator $A-B$ corresponding to non-negative eigenvalues, and set $\{A<B\}:=\mathbb{1}-\{A \geq B\}$. We further define $A_{+}:=\{A \geq 0\} A\{A \geq 0\}$ and take note of the following property:
Lemma 2.1 ([18]). For operators $A, B \geq 0$ and $0 \leq P \leq \mathbb{1}$ we have

$$
\operatorname{Tr}(A-B)_{+}=\operatorname{Tr}[\{A \geq B\}(A-B)] \geq \operatorname{Tr}[P(A-B)] .
$$

The inverse of the cumulative distribution function (c.d.f.) of a standard normal random variable is defined by

$$
\begin{equation*}
\Phi^{-1}(\varepsilon):=\sup \{z \in \mathbb{R} \mid \Phi(z) \leq \varepsilon\} \tag{2.1}
\end{equation*}
$$

where $\Phi(z)=\frac{1}{\sqrt{2 \pi}} \int_{-\infty}^{z} e^{-t^{2} / 2} d t$. Note that $\Phi(x)=1-\Phi(-x)$ and $\Phi^{-1}(1-x)=-\Phi^{-1}(x)$.
Two central quantities in our discussion are the quantum relative entropy $D(\rho \| \tau)$ and the quantum information variance $V(\rho \| \tau)$ :

Definition 2.2. Let $\rho \in \mathcal{D}(\mathcal{H})$ and $\tau \in \mathcal{P}(\mathcal{H})$.
(i) [27] The quantum relative entropy $D(\rho \| \tau)$ is defined as

$$
D(\rho \| \tau):= \begin{cases}\operatorname{Tr}[\rho(\log \rho-\log \tau)] & \text { if } \operatorname{supp} \rho \subseteq \operatorname{supp} \tau \\ \infty & \text { else. }\end{cases}
$$

Note that the von Neumann entropy is given by $S(\rho)=-D(\rho \| \mathbb{1})$.
(ii) [24] The quantum information variance $V(\rho \| \tau)$ is defined as

$$
V(\rho \| \tau):=\operatorname{Tr}\left[\rho(\log \rho-\log \tau)^{2}\right]-D(\rho \| \tau)^{2}
$$

Further, we define $\sigma(\rho \| \tau):=\sqrt{V(\rho \| \tau)}$ and

$$
\begin{equation*}
\sigma(\rho):=\sigma(\rho \| \mathbb{1})=\sqrt{V(\rho \| \mathbb{1})} \tag{2.2}
\end{equation*}
$$

Note that $\sigma(\rho)$ is equal to the standard deviation of the probability distribution formed by the eigenvalues of $\rho$. In the classical literature, the information variance of a source is sometimes also referred to as varentropy.

In [24] the authors introduced the information spectrum relative entropy $D_{s}^{\varepsilon}(\rho \| \tau)$, defined for $\varepsilon \in(0,1), \rho \in \mathcal{D}(\mathcal{H})$, and $\tau \in \mathcal{P}(\mathcal{H})$ as

$$
\begin{equation*}
D_{s}^{\varepsilon}(\rho \| \tau):=\sup \left\{\gamma \in \mathbb{R} \mid \operatorname{Tr}\left[\rho\left\{\rho \leq 2^{\gamma} \tau\right\}\right] \leq \varepsilon\right\} \tag{2.3}
\end{equation*}
$$

This quantity is particularly useful because its second order asymptotic expansion can be employed to obtain the second order asymptotics of quantum hypothesis testing, as shown in [24]. The derivation of our main results is based on the second order asymptotic expansion of the information spectrum relative entropy, which we employ in the following form:

Theorem $2.3([24])$. Let $\rho \in \mathcal{D}(\mathcal{H})$ with $S=S(\rho)$ and $\sigma=\sigma(\rho)$. There is a $K>0$ such that for any $L \in \mathbb{R}$ and $n \in \mathbb{N}$ we have

$$
\begin{equation*}
\left|\operatorname{Tr}\left(\rho^{\otimes n}\left\{\rho^{\otimes n} \leq 2^{-n S+\sqrt{n} L} \mathbb{1}\right\}\right)-\Phi\left(\frac{L}{\sigma}\right)\right| \leq \frac{K}{\sqrt{n}} \tag{2.4}
\end{equation*}
$$

Note however, that the trace expression on the left-hand side of (2.4) only depends on the eigenvalues of $\rho^{\otimes n}$. Hence, Theorem 2.3 already follows from the second order asymptotics of classical source coding derived by Strassen [22].

## 3 Operational setting

### 3.1 Mixed quantum sources

A general quantum information source is characterized by an ensemble $\mathfrak{E}=\left\{p_{i},\left|\psi_{i}\right\rangle\right\}_{i}$ of pure states (or signals) $\left|\psi_{i}\right\rangle \in \mathcal{H}$ which are emitted by the source with corresponding probabilities $p_{i}$. We refer to $\mathfrak{E}$ as the source ensemble, and the associated density matrix (or ensemble average state) $\rho=\sum_{i} p_{i} \psi_{i}$ is called the source state. A source is called memoryless if there
are no correlations between successive signals emitted by the source. Consequently, we can characterize $n$ uses of a memoryless source $\mathfrak{E}$ by the source ensemble $\mathfrak{E}^{n}=\left\{p_{\underline{i}},\left|\psi_{\underline{i}}\right\rangle\right\}_{\underline{i}}$ where $\underline{i}:=i_{1} i_{2} \ldots i_{n}$ is a sequence of indices of length $n$, and we define

$$
\begin{equation*}
p_{\underline{i}}:=p_{i_{1}} p_{i_{2}} \ldots p_{i_{n}} \quad \text { and } \quad\left|\psi_{\underline{i}}\right\rangle:=\left|\psi_{i_{1}}\right\rangle \otimes\left|\psi_{i_{2}}\right\rangle \otimes \ldots\left|\psi_{i_{n}}\right\rangle . \tag{3.1}
\end{equation*}
$$

The corresponding source state for $n$ uses of the source $\mathfrak{E}$ is given by $\rho^{\otimes n}$.
We now construct a mixed source consisting of memoryless sources. To this end, let $\Lambda$ be an arbitrary parameter space with a normalized measure $\mu$, i.e. $\int_{\Lambda} d \mu(\lambda)=1$. Consider a family of memoryless sources parametrized by $\lambda \in \Lambda$, with source ensemble $\mathfrak{E}_{\lambda}=\left\{q_{i}^{(\lambda)},\left|\varphi_{i}^{(\lambda)}\right\rangle\right\}_{i}$ and source state $\rho_{\lambda}=\sum_{i} q_{i}^{(\lambda)} \varphi_{i}^{(\lambda)}$. The mixed source is the one that emits all successive signals from the memoryless source $\mathfrak{E}_{\lambda}$ according to the probability measure $d \mu(\lambda)$. We denote the mixed source obtained from this construction by $\left(\rho_{\lambda}, d \mu(\lambda)\right)_{\lambda \in \Lambda}$. The source state $\rho^{(n)}$ for $n$ uses of $\left(\rho_{\lambda}, d \mu(\lambda)\right)_{\lambda \in \Lambda}$ is given by

$$
\begin{equation*}
\rho^{(n)}=\int_{\Lambda} \rho_{\lambda}^{\otimes n} d \mu(\lambda) \tag{3.2}
\end{equation*}
$$

and the corresponding (not necessarily finite) ensemble is given by

$$
\begin{equation*}
\mathfrak{E}_{\text {mix }}^{(n)}:=\left\{d \mu(\lambda) q_{\underline{i_{\lambda}}}^{(\lambda)} ;\left|\varphi_{\underline{i_{\lambda}}}^{(\lambda)}\right\rangle\right\}_{\underline{i_{\lambda}}, \lambda \in \Lambda}, \tag{3.3}
\end{equation*}
$$

where $\underline{i_{\lambda}}$ is a sequence of indices of length $n$ and $\left|\varphi_{\underline{i_{\lambda}}}^{(\lambda)}\right\rangle$ is a tensor product of $n$ pure states as in (3.1) for each $\lambda \in \Lambda$.

Let us consider the special case where the measure $\mu$ has finite support on points $\lambda_{1}, \ldots, \lambda_{k} \in$ $\Lambda$, corresponding to a discrete probability distribution $\left\{t_{j}\right\}_{j=1}^{k}$. Hence, we have $k$ memoryless quantum information sources with source ensembles $\mathfrak{E}_{j}=\left\{q_{i}^{(j)},\left|\varphi_{i}^{(j)}\right\rangle\right\}$ and source states $\rho_{j}=$ $\sum_{i} q_{i}^{(j)} \varphi_{i}^{(j)}$ for $j=1, \ldots, k$. The underlying source ensemble for $n$ uses of this mixed source is

$$
\begin{equation*}
\mathfrak{E}_{\text {mix }}^{(n)}:=\left\{t_{1} q_{\underline{i_{1}}}^{(1)}, \ldots, t_{k} q_{\underline{i_{k}}}^{(k)} ;\left|\varphi_{\underline{i_{1}}}^{(1)}\right\rangle, \ldots,\left|\varphi_{\underline{i_{\underline{k}}}}^{(k)}\right\rangle\right\}_{i_{\underline{j}}}, j=1, \ldots, k, \tag{3.4}
\end{equation*}
$$

and the source state is given by

$$
\begin{equation*}
\rho^{(n)}:=\sum_{j=1}^{k} t_{j} \rho_{j}^{\otimes n} . \tag{3.5}
\end{equation*}
$$

We denote such a discrete mixed source consisting of $k$ memoryless sources $\rho_{1}, \ldots, \rho_{k}$ by the tuple $\left(\left\{\rho_{j}\right\}_{j=1}^{k},\left\{t_{j}\right\}_{j=1}^{k}\right)$ or simply $\left(\rho_{j}, t_{j}\right)_{j=1}^{k}$. In the special case of two memoryless sources, $k=$ 2 , we set $t \equiv t_{1}$ (such that $\left.t_{2}=1-t\right)$ and write $\left(\rho_{1}, \rho_{2}, t\right)$ for the mixed source $\rho=t \rho_{1}+(1-t) \rho_{2}$. The parameter $t$ is also referred to as mixing parameter.

Finally, we also mention the special case of a mixed source where we have a fixed set of pure states $\left\{\left|\varphi_{i}\right\rangle\right\}_{i}$, and for $\lambda \in \Lambda$ the source $\mathfrak{E}_{\lambda}$ corresponds to a probability distribution $\left\{q_{i}^{(\lambda)}\right\}_{i}$ over the pure states $\left\{\left|\varphi_{i}\right\rangle\right\}_{i}$. That is, in this case we have $\left\{\left|\varphi_{i}^{(\lambda)}\right\rangle\right\}_{i}=\left\{\left|\varphi_{i}\right\rangle\right\}_{i}$ for all $\lambda \in \Lambda$. The source state $\rho_{\lambda}$ of the memoryless source $\mathfrak{E}_{\lambda}$ is then given by $\rho_{\lambda}=\sum_{i} q_{i}^{(\lambda)}\left|\varphi_{i}\right\rangle\left\langle\varphi_{i}\right|$.

### 3.2 Quantum source coding

In fixed-length quantum source coding the aim is to store the information emitted by the source in a compressed state $\rho_{c} \in \mathcal{D}\left(\mathcal{H}_{c}\right)$ with $\operatorname{dim} \mathcal{H}_{c}<\operatorname{dim} \mathcal{H}$, such that it can later be decompressed
yielding a state which is sufficiently close to the source state $\rho$ with respect to some chosen distance measure.

There are two different settings $[1,10,29]$ for the compression part of the protocol outlined above: visible and blind. In this paper we only consider the visible setting. ${ }^{2}$ In this setting, the compressor (say, Alice) knows the identity of the signals $\psi_{i}$. In fact, on each use of the source Alice receives classical information in the form of an index $i$ labelling the signal $\psi_{i}$ emitted by the source. She then uses an arbitrary map $\mathcal{V}:\{i\} \rightarrow \mathcal{D}\left(\mathcal{H}_{c}\right)$ to encode the signal $\psi_{i}$ in a state $\mathcal{V}(i) \in \mathcal{D}\left(\mathcal{H}_{c}\right)$. We stress that $\mathcal{V}$ (which we refer to as visible encoding) is not a CPTP map acting on the signals $\psi_{i}$; Alice simply prepares a quantum state $\mathcal{V}(i)$ on receiving the index $i$. This is in contrast to the blind setting of source coding, where the encoder does not have any knowledge about the pure states $\psi_{i}$ and is therefore required to apply a quantum operation $\mathcal{E}$ to the source state $\rho$. Henceforth, we restrict the discussion to the visible setting. In the decompression part of the protocol, the compressed signal $\mathcal{V}(i)$ is subjected to a quantum operation $\mathfrak{D}: \mathcal{D}\left(\mathcal{H}_{c}\right) \rightarrow \mathcal{D}(\mathcal{H})$ which we call the decoding map.

### 3.3 Definition of the second order asymptotic rate

Our aim is to derive the second order asymptotic rate (or in short, second order rate) for fixedlength visible quantum source coding of a mixed source, whose precise definition we give below. Since we only discuss the visible source coding setting in this paper, we will henceforth suppress the attribute 'visible' in all definitions.

We choose the ensemble average fidelity as the figure of merit in our analysis of fixed-length quantum source coding, defined as follows:

Definition 3.1. Let $\mathfrak{E}=\left\{p_{i},\left|\psi_{i}\right\rangle\right\}_{i}$ be a pure-state ensemble with $\left|\psi_{i}\right\rangle \in \mathcal{H}$ for all $i$. We say that the triple $\mathcal{C}=(\mathcal{V}, \mathfrak{D}, M)$ defines a code for fixed-length visible source coding if $\mathcal{V}:\{i\} \rightarrow \mathcal{D}\left(\mathcal{H}_{c}\right)$ is an arbitrary encoding map, $\mathfrak{D}: \mathcal{D}\left(\mathcal{H}_{c}\right) \rightarrow \mathcal{D}(\mathcal{H})$ is a decoding CPTP map, and $\mathcal{H}_{c}$ is the compressed Hilbert space with $M:=\operatorname{dim} \mathcal{H}_{c}<\mathcal{H}$.

The ensemble average fidelity $\bar{F}(\mathfrak{E}, \mathcal{C})$ of the ensemble $\mathfrak{E}$ and the code $\mathcal{C}$ is defined as

$$
\bar{F}(\mathfrak{E}, \mathcal{C}):=\sum_{i} p_{i} \operatorname{Tr}\left((\mathfrak{D} \circ \mathcal{V})(i) \psi_{i}\right)
$$

For a mixed source $\left(\rho_{\lambda}, d \mu(\lambda)\right)_{\lambda \in \Lambda}$ as defined in Section 3.1 with ensemble $\mathfrak{E}_{\text {mix }}$ given as in (3.3) for $n=1$, the ensemble average fidelity $\bar{F}\left(\mathfrak{E}_{\text {mix }}, \mathcal{C}\right)$ is correspondingly defined as

$$
\bar{F}\left(\mathfrak{E}_{\mathrm{mix}}, \mathcal{C}\right):=\int_{\lambda \in \Lambda} d \mu(\lambda) \sum_{i} q_{i}^{(\lambda)} \operatorname{Tr}\left((\mathfrak{D} \circ \mathcal{V})(i) \psi_{i}^{(\lambda)}\right)
$$

This leads to the following definition:
Definition 3.2. Let $\left(\rho_{\lambda}, d \mu(\lambda)\right)_{\lambda \in \Lambda}$ be a mixed source, and let $\varepsilon \in(0,1)$. For $n \in \mathbb{N}$ let $\mathfrak{E}_{\text {mix }}^{(n)}$ as defined in (3.3) be the source ensemble for $n$ uses of the mixed source $\left(\rho_{\lambda}, d \mu(\lambda)\right)_{\lambda \in \Lambda}$. Given $R \in \mathbb{R}$, we say that any $r \in \mathbb{R}$ is an $(R, \varepsilon)$-achievable rate if there exists a sequence $\left\{\mathcal{C}_{n}\right\}_{n \in \mathbb{N}}$ of codes $\mathcal{C}_{n}=\left(\mathcal{V}_{n}, \mathfrak{D}_{n}, M_{n}\right)$ such that

$$
\begin{equation*}
\liminf _{n \rightarrow \infty} \bar{F}\left(\mathfrak{E}_{\mathrm{mix}}^{(n)}, \mathcal{C}_{n}\right) \geq 1-\varepsilon \quad \text { and } \quad \limsup _{n \rightarrow \infty} \frac{\log M_{n}-n R}{\sqrt{n}} \leq r \tag{3.6}
\end{equation*}
$$

The second order asymptotic rate $b(R, \varepsilon \mid \rho)$ for $n$ uses of the mixed source $\left(\rho_{\lambda}, d \mu(\lambda)\right)_{\lambda \in \Lambda}$ is then defined as the infimum over all $(R, \varepsilon)$-achievable rates $r$.

[^2]
## Remark 3.3.

(i) For any $R>0$ the quantity $b(R, \varepsilon \mid \rho)$ is only finite if the parameter $R$ equals the optimal first order rate $a$ of the protocol, i.e. a real number $a$ satisfying

$$
\begin{equation*}
\log M_{n}=n a+f(n) \tag{3.7}
\end{equation*}
$$

with $f(n) \in \mathcal{O}(\sqrt{n})$. This can be seen as follows: Substituting (3.7) in (3.6) of Definition 3.2(ii) yields

$$
\begin{equation*}
\frac{n a-n R}{\sqrt{n}}+\frac{f(n)}{\sqrt{n}}=\sqrt{n}(a-R)+\frac{f(n)}{\sqrt{n}} \tag{3.8}
\end{equation*}
$$

Taking the limit superior in (3.8), the second term is some constant since $f(n) \in \mathcal{O}(\sqrt{n})$, whereas the first term diverges to either $+\infty$ if $R<a$ or $-\infty$ if $R>a$.
(ii) For quantum source coding using a single memoryless source, $a$ is equal to the von Neumann entropy $S(\rho)$ of the source, and (3.7) is proven in [20, 28].

## 4 Main results

Our main result is the derivation of the second order asymptotic rate for $n$ uses of a mixed source $\left(\rho_{\lambda}, d \mu(\lambda)\right)_{\lambda \in \Lambda}$ with source state $\rho^{(n)}=\int_{\Lambda} \rho_{\lambda}^{\otimes n} d \mu(\lambda)$ as defined in Section 3.1. In order to state our main result, we make the following definition: For a fixed $a>0$ let $\mathcal{L}_{=}(a):=\{\lambda \in$ $\left.\Lambda: S\left(\rho_{\lambda}\right)=a\right\}$ and $\mathcal{L}_{<}(a):=\left\{\lambda \in \Lambda: S\left(\rho_{\lambda}\right)<a\right\}$. Furthermore, recall that for $\rho \in \mathcal{D}(\mathcal{H})$ we set $\sigma(\rho)=\sqrt{V(\rho \| \mathbb{1})}$ (cf. (2.2)). We then have:

Theorem 4.1. Let $\Lambda$ be an arbitrary parameter space with a normalized measure $\mu$, that is, $\int_{\Lambda} d \mu(\lambda)=1$, and let $\left(\rho_{\lambda}, d \mu(\lambda)\right)_{\lambda \in \Lambda}$ be a mixed source. Furthermore, let $a>0, \varepsilon \in(0,1)$, and define $\sigma_{\lambda}=\sigma\left(\rho_{\lambda}\right)$ for $\lambda \in \Lambda$. Then the second order asymptotic rate $b(a, \varepsilon \mid \rho)$ for $n$ uses of the mixed source $\left(\rho_{\lambda}, d \mu(\lambda)\right)_{\lambda \in \Lambda}$ is the solution of the equation

$$
\int_{\mathcal{L}_{=}(a)} \Phi\left(\frac{b}{\sigma_{\lambda}}\right) d \mu(\lambda)+\int_{\mathcal{L}_{<}(a)} d \mu(\lambda)=1-\varepsilon
$$

If the measure $\mu$ has finite support on points $\lambda_{1}, \ldots, \lambda_{k} \in \Lambda$, Theorem 4.1 reduces to the following
Corollary 4.2. Consider a mixed source $\rho=\left(\rho_{j}, t_{j}\right)_{j=1}^{k}$, and set $S_{j}=S\left(\rho_{j}\right)$ and $\sigma_{j}=\sigma\left(\rho_{j}\right)$ for $j=1, \ldots, k$. For $a>0$ and $\varepsilon \in(0,1)$, the second order asymptotic rate $b(a, \varepsilon \mid \rho)$ for $n$ uses of the mixed source $\rho=\left(\rho_{j}, t_{j}\right)_{j=1}^{k}$ is given by the solution of the equation

$$
\sum_{i: S_{i}=a} t_{i} \Phi\left(\frac{L}{\sigma_{i}}\right)+\sum_{i: S_{i}<a} t_{i}=1-\varepsilon
$$

Finally, we consider the special case of a mixed source consisting of two memoryless sources $\rho_{1}, \rho_{2} \in \mathcal{D}(\mathcal{H})$ with corresponding source state

$$
\rho^{(n)}=t \rho_{1}^{\otimes n}+(1-t) \rho_{2}^{\otimes n}
$$

and mixing parameter $t \in(0,1)$. We adhere to the discussion of classical mixed source coding by Nomura and Han [16] by considering the following three cases, ${ }^{3}$ abbreviating $S_{i} \equiv S\left(\rho_{i}\right)$ and $\sigma_{i} \equiv \sigma\left(\rho_{i}\right)$ for $i=1,2$ :

[^3]Case 1: $S_{1}=S_{2}$
Case 2: $S_{1}>S_{2}, t>\varepsilon$
Case 3: $S_{1}>S_{2}, t<\varepsilon$
We state the second order rate in each of the three cases in the following theorem:
Theorem 4.3. Consider a mixed source $\rho=\left(\rho_{1}, \rho_{2}, t\right)$ with $\rho_{1}, \rho_{2} \in \mathcal{D}(\mathcal{H})$ and $t \in(0,1)$, and set $S_{i}:=S\left(\rho_{i}\right)$ and $\sigma_{i}:=\sigma\left(\rho_{i}\right)$ for $i=1,2$. For $\varepsilon \in(0,1)$ the second order asymptotic rate $b(a, \varepsilon \mid \rho)$ for $n$ uses of the mixed source $\left(\rho_{1}, \rho_{2}, t\right)$ is given by the following expressions:
(i) For $S_{1}=S_{2} \equiv S$, we have $b(S, \varepsilon \mid \rho)=L$ where $L$ is the solution of the equation

$$
\begin{equation*}
t \Phi\left(\frac{L}{\sigma_{1}}\right)+(1-t) \Phi\left(\frac{L}{\sigma_{2}}\right)=1-\varepsilon \tag{4.1}
\end{equation*}
$$

(ii) For $S_{1}>S_{2}$ and $t>\varepsilon$, we have

$$
\begin{equation*}
b\left(S_{1}, \varepsilon \mid \rho\right)=-\sigma_{1} \Phi^{-1}\left(\frac{\varepsilon}{t}\right) \tag{4.2}
\end{equation*}
$$

(iii) For $S_{1}>S_{2}$ and $t<\varepsilon$, we have

$$
\begin{equation*}
b\left(S_{2}, \varepsilon \mid \rho\right)=-\sigma_{2} \Phi^{-1}\left(\frac{\varepsilon-t}{1-t}\right) \tag{4.3}
\end{equation*}
$$

## Remark 4.4.

(i) Upon replacing the quantum sources $\rho_{\lambda}$ with classical i.i.d. sources characterized by a random variable $Y_{\lambda}$, identifying $S\left(\rho_{\lambda}\right)$ with the Shannon entropy $H\left(Y_{\lambda}\right)$, and the quantum information variance $\sigma_{\lambda}$ with the standard deviation of the random variable $\log Y_{\lambda}$, Theorem 4.1 and Theorem 4.3 reproduce Theorem 8.3 and Theorem 7.1 in [16], respectively, in the case of a finite source alphabet.
(ii) Recall from Remark 3.3(i) that the statement $b\left(S_{1}, \varepsilon \mid \rho\right)=-\sigma_{1} \Phi^{-1}(\varepsilon / t)<\infty$ in Theorem 4.3(ii) implies that the first order rate equals $S_{1}$. In particular, in this case $b\left(S_{2}, \varepsilon \mid \rho\right)=$ $\infty$. Similarly, in Theorem 4.3(iii) the first order rate is given by $S_{2}$, and $b\left(S_{1}, \varepsilon \mid \rho\right)=-\infty$.
(iii) To determine the range of $L$ in Theorem 4.3(i), assume without loss of generality that $\sigma_{1}<\sigma_{2}$. Then, using properties of the c.d.f. $\Phi$ of a normal distribution and definition (4.1) of $L$, it follows easily that

$$
\begin{array}{ll}
L \in\left[-\sigma_{1} \Phi^{-1}(\varepsilon),-\sigma_{2} \Phi^{-1}(\varepsilon)\right] & \text { if } \varepsilon \in(0,1 / 2) \\
L \in\left[-\sigma_{2} \Phi^{-1}(\varepsilon),-\sigma_{1} \Phi^{-1}(\varepsilon)\right] & \text { if } \varepsilon \in(1 / 2,1) \tag{4.4b}
\end{array}
$$

and $L=0$ for $\varepsilon=1 / 2$. See Figure 1 for a plot showing a typical example of this.


Figure 1: Plot of the second order asymptotic rate $L$ (blue-solid) defined in (4.1) and bounds on $L$ (red-dashed and green-dash-dotted) for $\varepsilon \in(0,1 / 2)$ (4.4a) and $\varepsilon \in(1 / 2,1)$ (4.4b) for a mixed source ( $\rho_{1}, \rho_{2}, t$ ) with the values $\sigma_{1}=0.235, \sigma_{2}=0.712$, and $t=0.425$.

## 5 Proofs

The following lemma is a direct consequence of Theorem 2.3 and a key ingredient in the proof of Theorem 4.1.

Lemma 5.1. Let $\rho_{1}, \rho_{2} \in \mathcal{D}(\mathcal{H})$ with $S_{i}:=S\left(\rho_{i}\right)$ and $\sigma_{i}:=\sigma\left(\rho_{i}\right)$ for $i=1,2$. If $S_{1}>S_{2}$, then for any constant $C$ we have:

$$
\begin{align*}
& \lim _{n \rightarrow \infty} \operatorname{Tr}\left(\rho_{2}^{\otimes n}\left\{\rho_{2}^{\otimes n} \leq 2^{-n S_{1}-\sqrt{n} C} \mathbb{1}\right\}\right)=0  \tag{5.1}\\
& \lim _{n \rightarrow \infty} \operatorname{Tr}\left(\rho_{1}^{\otimes n}\left\{\rho_{1}^{\otimes n} \leq 2^{-n S_{2}-\sqrt{n} C} \mathbb{1}\right\}\right)=1 \tag{5.2}
\end{align*}
$$

Proof. In order to prove (5.1), define $f_{n}:=\sqrt{n}\left(S_{1}-S_{2}\right)$ and note that $f_{n} \xrightarrow{n \rightarrow \infty} \infty$ by assumption. We then obtain the following bound for some constant $K>0$ :

$$
\begin{aligned}
\operatorname{Tr}\left(\rho_{2}^{\otimes n}\left\{\rho_{2}^{\otimes n} \leq 2^{-n S_{1}-\sqrt{n} C_{1}}\right\}\right) & =\operatorname{Tr}\left(\rho_{2}^{\otimes n}\left\{\rho_{2}^{\otimes n} \leq 2^{-n S_{2}-\sqrt{n}\left(C+f_{n}\right)} \mathbb{1}\right\}\right) \\
& \leq \Phi\left(-\frac{C+f_{n}}{\sigma_{2}}\right)+\frac{K}{\sqrt{n}}
\end{aligned}
$$

where the inequality follows from Theorem 2.3. This yields (5.1) since $\lim _{x \rightarrow-\infty} \Phi(x)=0$. Identity (5.2) is proved along similar lines.

We also state the following result by Hayashi [10], which gives an upper bound on the ensemble average fidelity. For a proof in our notation, see Proposition 7 in Section V.A of [7].

Lemma 5.2 ([10]). Let $\mathfrak{E}=\left\{p_{i}, \psi_{i}\right\}_{i}$ be an ensemble of pure states and set $\rho=\sum_{i} p_{i} \psi_{i}$. Let $\mathcal{V}:\{i\} \rightarrow \mathcal{D}\left(\mathcal{H}_{c}\right)$ be a visible encoding map with $\mathcal{H}_{c}$ denoting the compressed Hilbert space with $\operatorname{dim} \mathcal{H}_{c}=M$, and let $\mathfrak{D}: \mathcal{D}\left(\mathcal{H}_{c}\right) \rightarrow \mathcal{D}(\mathcal{H})$ denote the decoding CPTP map. Then for the code $\mathcal{C}=(\mathcal{V}, \mathfrak{D}, M)$ we have

$$
\bar{F}(\mathfrak{E}, \mathcal{C}) \leq \max \{\operatorname{Tr}(P \rho) \mid P \text { is a projection on } \mathcal{H} \text { with } \operatorname{Tr} P=M\}
$$

We can now prove an upper bound on the ensemble average fidelity that we need for proving the converse bounds of Theorem 4.1 and Theorem 4.3.

Lemma 5.3. Let $\left(\left\{\rho_{j}\right\}_{j=1}^{k},\left\{t_{j}\right\}_{j=1}^{k}\right)$ be a mixed source with corresponding source state $\rho=$ $\sum_{j=1}^{k} t_{j} \rho_{j}$ and ensemble $\mathfrak{E}_{\text {mix }}$ defined in (3.4) for $n=1$. For any code $\mathcal{C}=(\mathcal{V}, \mathfrak{D}, M)$ and $\gamma \in \mathbb{R}$, the ensemble average fidelity satisfies

$$
\bar{F}\left(\mathfrak{E}_{\text {mix }}, \mathcal{C}\right) \leq 1-\sum_{j=1}^{k} t_{j} \operatorname{Tr}\left(\rho_{j}\left\{\rho_{j} \leq 2^{-\gamma} \mathbb{1}\right\}\right)+2^{-\gamma+\log M}
$$

Proof. By Lemma 5.2 there is a projection $Q$ with $\operatorname{Tr} Q=M$ such that $\bar{F}\left(\mathfrak{E}_{\text {mix }}, \mathcal{C}\right) \leq \operatorname{Tr}(Q \rho)$. For arbitrary $\gamma \in \mathbb{R}$, we then compute:

$$
\begin{aligned}
\bar{F}\left(\mathfrak{E}_{\text {mix }}, \mathcal{C}\right) \leq & \operatorname{Tr}(Q \rho) \\
= & \sum_{j=1}^{k} t_{j} \operatorname{Tr} Q \rho_{j} \\
= & \sum_{j=1}^{k} t_{j} \operatorname{Tr}\left[Q\left(\rho_{j}-2^{-\gamma} \mathbb{1}\right)\right]+2^{-\gamma} \operatorname{Tr} Q \\
\leq & \sum_{j=1}^{k} t_{j} \operatorname{Tr}\left[\left\{\rho_{j}>2^{-\gamma} \mathbb{1}_{n}\right\}\left(\rho_{j}-2^{-\gamma} \mathbb{1}\right)\right]+2^{-\gamma+\log M} \\
= & 1-2^{-\gamma} \operatorname{Tr} \mathbb{1}-\sum_{j=1}^{k} t_{j} \operatorname{Tr}\left(\left\{\rho_{j} \leq 2^{-\gamma} \mathbb{1}\right\}\left(\rho_{j}-2^{-\gamma} \mathbb{1}\right)\right)+2^{-\gamma+\log M} \\
= & 1-2^{-\gamma} \operatorname{Tr} \mathbb{1}-\sum_{j=1}^{k} t_{j} \operatorname{Tr}\left(\rho_{j}\left\{\rho_{j} \leq 2^{-\gamma} \mathbb{1}\right\}\right) \\
& +2^{-\gamma} \sum_{j=1}^{k} t_{j} \operatorname{Tr}\left\{\rho_{j} \leq 2^{-\gamma} \mathbb{1}\right\}+2^{-\gamma+\log M} \\
\leq & 1-\sum_{j=1}^{k} t_{j} \operatorname{Tr}\left(\rho_{j}\left\{\rho_{j} \leq 2^{-\gamma} \mathbb{1}\right\}\right)+2^{-\gamma+\log M}
\end{aligned}
$$

where we used Lemma 2.1 in the second inequality, the identity $\left\{\rho_{j}>2^{-\gamma} \mathbb{1}\right\}=\mathbb{1}-\left\{\rho_{j} \leq 2^{-\gamma} \mathbb{1}\right\}$ in the third equality, and $\left\{\rho_{j} \leq 2^{-\gamma} \mathbb{1}\right\} \leq \mathbb{1}$ in the last inequality.

We also record the following simple observation: Let $A, B, C \in \mathcal{P}(\mathcal{H})$ be pairwise commuting operators with $B \leq C$. Then we have $\{A \leq B\} \leq\{A \leq C\}$, which can easily be seen to be true by considering a common eigenbasis of $A, B$, and $C$ and checking the corresponding relation in the scalar case. We will use this result in the following form:

Lemma 5.4. Let $a, b \in \mathbb{R}$ with $a \leq b$, then for any $X \geq 0$ we have

$$
\left\{X \leq 2^{-b} \mathbb{1}\right\} \leq\left\{X \leq 2^{-a} \mathbb{1}\right\}
$$

For the remainder of this section, we abbreviate $\rho^{n} \equiv \rho^{\otimes n}$.

### 5.1 Universal source code achieving second order asymptotic rates

In this section we construct a universal source code that, given parameters $a \in \mathbb{R}$ (which is to be chosen later as the first order rate) and $\varepsilon \in(0,1)$, achieves a second order asymptotic rate $b(a, \varepsilon \mid \rho)$ for any $\rho \in \mathcal{D}(\mathcal{H})$. Our construction relies on ideas taken from papers by Jozsa et al. [12] and Hayashi [11].

Let $\mathcal{X}=\{1, \ldots, d\}$. The type $P_{\underline{x}}$ of a sequence $\underline{x}=x_{1} \ldots x_{n} \in \mathcal{X}^{n}$ is the empirical distribution of the letters of $\mathcal{X}$ in $\underline{x}$, that is, $P_{\underline{x}}(x)=\frac{1}{n} \sum_{i=1}^{n} \delta_{x_{i}, x}$ for all $x \in \mathcal{X}$. We denote by $\mathcal{T}_{n}$ the set of all types, and for a type $P \in \mathcal{T}_{n}$ we denote by $T_{P}^{n} \subset \mathcal{X}^{n}$ the set of sequences of type $P$. Following [11], for $a, b \in \mathbb{R}$ we define

$$
T_{n}(a, b):=\bigcup\left\{T_{P}^{n} \mid P \in \mathcal{T}_{n} \text { with }\left|T_{P}^{n}\right| \leq \exp (a n+b \sqrt{n})\right\} \subset \mathcal{X}^{n}
$$

A simple type-counting argument [5] shows that

$$
\left|T_{n}(a, b)\right| \leq(n+1)^{d} \exp (a n+b \sqrt{n}) .
$$

Let now $B=\left\{\left|e_{1}\right\rangle, \ldots,\left|e_{d}\right\rangle\right\}$ be a basis of $\mathcal{H}$. As in [12], we define the subspace

$$
\Xi_{a, b}^{n}(B):=\operatorname{span}\left\{\left|e_{\underline{i}}\right\rangle \in B^{\otimes n} \mid \underline{i} \in T_{n}(a, b)\right\},
$$

that is, $\Xi_{a, b}^{n}(B)$ is the span of basis vectors of the product basis $B^{\otimes n}$ of $\mathcal{H}^{\otimes n}$ labelled by sequences in $T_{n}(a, b)$. The code space $\Upsilon_{a, b}^{n}$ of the universal source code is now obtained by varying $B$ over all bases of $\mathcal{H}$. More precisely, we define $\Upsilon_{a, b}^{n}$ as the smallest subspace of $\mathcal{H}^{\otimes n}$ containing $\Xi_{a, b}^{n}(B)$ for all bases $B$ of $\mathcal{H}$. To estimate the size of $\Upsilon_{a, b}^{n}$, we use the following

Lemma 5.5 ([12]). Let $|\phi\rangle \in \mathcal{H}^{\otimes n}$ with $\operatorname{dim} \mathcal{H}=d$, and let $\mathcal{H}_{\phi}:=\operatorname{span}\left\{A^{\otimes n}|\phi\rangle \mid A \in \mathcal{B}(\mathcal{H})\right\}$, then $\operatorname{dim} \mathcal{H}_{\phi} \leq(n+1)^{d^{2}}$.

We now obtain:
Lemma 5.6. With the above definitions, the dimension of the code space $\Upsilon_{a, b}^{n} \subseteq \mathcal{H}^{\otimes n}$ can be estimated as

$$
\operatorname{dim} \Upsilon_{a, b}^{n} \leq(n+1)^{d^{2}+d} \exp (a n+b \sqrt{n}) .
$$

Proof. Here, we closely follow an argument in [12]. First, let $B_{0}$ be a fixed basis of $\mathcal{H}$. Then any other basis $B$ can be obtained from $B_{0}$ by applying some unitary operator $U$ on the basis vectors of $B_{0}$. As $\Xi_{a, b}^{n}(B)$ is the span of tensor products of elements in $B$, we have

$$
\left.\Xi_{a, b}^{n}(B)=\left\{U^{\otimes n}|\phi\rangle| | \phi\right\rangle \in \Xi_{a, b}^{n}\left(B_{0}\right)\right\} .
$$

Hence, the following holds for the code space $\Upsilon_{a, b}^{n}$ :

$$
\begin{aligned}
\Upsilon_{a, b}^{n} & \left.=\operatorname{span}\left\{U^{\otimes n}|\phi\rangle|U \in \mathcal{U}(d),| \phi\right\rangle \in \Xi_{a, b}^{n}\left(B_{0}\right)\right\} \\
& \left.\subset \operatorname{span}\left\{A^{\otimes n}|\phi\rangle|A \in \mathcal{B}(\mathcal{H}),| \phi\right\rangle \in \Xi_{a, b}^{n}\left(B_{0}\right)\right\}
\end{aligned}
$$

As $\operatorname{dim} \Xi_{a, b}^{n}\left(B_{0}\right) \leq\left|T_{n}(a, b)\right| \leq(n+1)^{d} \exp (a n+b \sqrt{n})$, the claim now follows from Lemma 5.5.

Proposition 5.7 (Universal code achieving second order rate). Let $\mathfrak{E}=\left\{p_{i}, \psi_{i}\right\}_{i}$ be the purestate ensemble of an arbitrary memoryless quantum source with associated source state $\rho \in$
$\mathcal{D}(\mathcal{H})$, and abbreviate $S \equiv S(\rho)$ and $\sigma \equiv \sigma(\rho)$. Let $\Pi_{n}$ be the projector onto the code space $\Upsilon_{S, b}^{n}$ defined as above, and consider the visible encoding map

$$
\begin{equation*}
\mathcal{V}_{n}: \underline{i} \longmapsto \frac{\Pi_{n} \psi_{\underline{i}} \Pi_{n}}{\operatorname{Tr}\left(\Pi_{n} \psi_{\underline{i}}\right)} \tag{5.3}
\end{equation*}
$$

We set $M_{n}:=\operatorname{dim} \Upsilon_{S, b}^{n}$, and define the decoding operation $\mathfrak{D}_{n}: \Upsilon_{S, b}^{n} \rightarrow \mathcal{H}^{\otimes n}$ as the trivial embedding. For $n$ uses of the source $\rho$, the sequence $\left\{\mathcal{C}_{n}\right\}_{n \in \mathbb{N}}$ of codes $\mathcal{C}_{n}=\left(\mathcal{V}_{n}, \mathfrak{D}_{n}, M_{n}\right)$ then achieves the second order rate $b=b(S, \varepsilon \mid \rho)$, where $\varepsilon=1-\Phi(b / \sigma)$.
Proof. Lemma 5.6 immediately yields

$$
\limsup _{n \rightarrow \infty} \frac{\log M_{n}-S n}{\sqrt{n}} \leq \limsup _{n \rightarrow \infty} \frac{\left(d^{2}+d\right) \log (n+1)}{\sqrt{n}}+b=b
$$

With the visible encoding given by (5.3), we can express the ensemble average fidelity $\bar{F}\left(\mathfrak{E}^{n}, \mathcal{C}_{n}\right)$ as [7, Sect. V.A.3]

$$
\begin{equation*}
\bar{F}\left(\mathfrak{E}^{n}, \mathcal{C}_{n}\right)=\operatorname{Tr}\left(\rho^{\otimes n} \Pi_{n}\right) \tag{5.4}
\end{equation*}
$$

We now employ the following relation proved by Hayashi [11] in the context of classical fixedlength source coding:

$$
\mathcal{S}_{n}:=\left\{\underline{x} \in \mathcal{X}^{n} \mid-\log P^{n}(\underline{x})<n a+\sqrt{n} b\right\} \subseteq T_{n}(a, b)
$$

which holds for arbitrary $a, b \in \mathbb{R}$ and probability distributions $P$ with support on $\{1, \ldots, d\}$. Consider the spectral decomposition $\rho=\sum_{i} r_{i}\left|\varphi_{i}\right\rangle\left\langle\varphi_{i}\right|$, and set $P_{\rho}=\left\{r_{i}\right\}_{i}$ and $B_{\rho}=\left\{\left|\varphi_{i}\right\rangle\right\}_{i}$. Observe that the projector $\left\{\rho^{\otimes n}>\exp (-n a-\sqrt{n} b) \mathbb{1}_{n}\right\}$ projects onto eigenvectors of $\rho^{\otimes n}$ labelled by elements of $\mathcal{S}_{n}$, upon choosing $P=P_{\rho}$. Since the code space $\Upsilon_{a, b}^{n}$ includes the subspace $\Xi_{a, b}^{n}\left(B_{\rho}\right)$, we have the operator inequality

$$
\begin{equation*}
\Pi_{n} \geq\left\{\rho^{\otimes n}>\exp (-n a-\sqrt{n} b) \mathbb{1}_{n}\right\} \tag{5.5}
\end{equation*}
$$

We now set $a=S$ in (5.5) and substitute it in (5.4). Taking the limit inferior, we obtain

$$
\begin{aligned}
\liminf _{n \rightarrow \infty} \bar{F}\left(\mathfrak{E}^{n}, \mathcal{C}_{n}\right) & =\liminf _{n \rightarrow \infty} \operatorname{Tr}\left(\rho^{\otimes n} \Pi_{n}\right) \\
& \geq \liminf _{n \rightarrow \infty} \operatorname{Tr}\left[\rho^{\otimes n}\left\{\rho^{\otimes n}>\exp (-n S-\sqrt{n} b) \mathbb{1}_{n}\right\}\right] \\
& =1-\limsup _{n \rightarrow \infty} \operatorname{Tr}\left[\rho^{\otimes n}\left\{\rho^{\otimes n} \leq \exp (-n S-\sqrt{n} b) \mathbb{1}_{n}\right\}\right] \\
& =1-\Phi\left(-\frac{b}{\sigma}\right) \\
& =\Phi\left(\frac{b}{\sigma}\right)
\end{aligned}
$$

where we used Theorem 2.3 in the third equality. Setting $\varepsilon:=1-\Phi(b / \sigma)$ now yields the claim.

### 5.2 General mixture

In this section we prove the assertion of Theorem 4.1, which states that for $a>0$ and $\varepsilon \in(0,1)$ the second order asymptotic rate $b(a, \varepsilon \mid \rho)$ for $n$ uses of a general mixed source $\left(\rho_{\lambda}, d \mu(\lambda)\right)_{\lambda \in \Lambda}$ with source state $\rho^{(n)}=\int_{\Lambda} \rho_{\lambda}^{\otimes n} d \mu(\lambda)$ is given by the solution of the relation

$$
\begin{equation*}
\int_{\mathcal{L}_{=(a)}} \Phi\left(\frac{b}{\sigma_{\lambda}}\right) d \mu(\lambda)+\int_{\mathcal{L}_{<}(a)} d \mu(\lambda)=1-\varepsilon \tag{5.6}
\end{equation*}
$$

Here, the sets $\mathcal{L}_{=}(a)$ and $\mathcal{L}_{<}(a)$ are defined by

$$
\begin{equation*}
\mathcal{L}_{=}(a):=\left\{\lambda \in \Lambda: S\left(\rho_{\lambda}\right)=a\right\} \quad \mathcal{L}_{<}(a):=\left\{\lambda \in \Lambda: S\left(\rho_{\lambda}\right)<a\right\} \tag{5.7}
\end{equation*}
$$

and we set $\sigma_{\lambda}:=\sigma\left(\rho_{\lambda}\right)$ (cf. Definition 2.2(ii)). Before we proceed with the proof, we note that the converse bound on the ensemble average fidelity in Lemma 5.3 holds for arbitrary ensembles $\left\{d \mu(\lambda), \psi_{\lambda}\right\}_{\lambda \in \Lambda}$ with respect to the measure $\mu$ on $\Lambda$. Here, $\psi_{\lambda} \in \mathcal{D}(\mathcal{H})$ is a pure state for $\lambda \in \Lambda$, and $\rho=\int_{\Lambda} \psi_{\lambda} d \mu(\lambda)$ is the corresponding ensemble average state.

### 5.2.1 Converse bound

Denoting the solution of (5.6) by $b^{*}$, we first prove the converse statement, i.e. $b(a, \varepsilon \mid \rho) \geq b^{*}$. To this end, assume that $R<b^{*}$ is an $(a, \varepsilon)$-achievable second order rate, that is, there is a sequence $\left\{\mathcal{C}_{n}\right\}_{n \in \mathbb{N}}$ of codes $\mathcal{C}_{n}=\left(\mathcal{V}_{n}, \mathfrak{D}_{n}, M_{n}\right)$ for $n$ uses of the mixed source $\left(\rho_{\lambda}, d \mu(\lambda)\right)_{\lambda \in \Lambda}$ with source ensemble $\mathfrak{E}_{\text {mix }}^{(n)}$ such that

$$
\begin{align*}
\liminf _{n \rightarrow \infty} \bar{F}\left(\mathfrak{E}_{\operatorname{mix}}^{(n)}, \mathcal{C}_{n}\right) & \geq 1-\varepsilon  \tag{5.8a}\\
\limsup _{n \rightarrow \infty} & \frac{\log M_{n}-n a}{\sqrt{n}} \tag{5.8b}
\end{align*}
$$

Choose $\delta>0$ such that $R+2 \delta<b^{*}$. Then by (5.8b) we have for sufficiently large $n$ that

$$
\begin{equation*}
\log M_{n}<n a+\sqrt{n}(R+\delta) \tag{5.9}
\end{equation*}
$$

Lemma 5.3 yields the following bound on the fidelity $\bar{F}\left(\mathfrak{E}_{\text {mix }}^{(n)}, \mathcal{C}_{n}\right)$ for arbitrary $\gamma \in \mathbb{R}$ :

$$
\bar{F}\left(\mathfrak{E}_{\text {mix }}^{(n)}, \mathcal{C}_{n}\right) \leq 1-\int_{\Lambda} \operatorname{Tr}\left(\rho_{\lambda}^{n}\left\{\rho_{\lambda}^{n} \leq 2^{-\gamma} \mathbb{1}_{n}\right\}\right) d \mu(\lambda)+2^{-\gamma+\log M_{n}}
$$

We now set $\gamma=\log M_{n}+\sqrt{n} \delta$, such that by (5.9) we have

$$
\gamma<n a+\sqrt{n}(R+2 \delta)
$$

Hence, Lemma 5.4 yields

$$
\begin{align*}
& \bar{F}\left(\mathfrak{E}_{\text {mix }}^{(n)}, \mathcal{C}_{n}\right) \leq 1-\int_{\Lambda} \operatorname{Tr}\left(\rho_{\lambda}^{n}\left\{\rho_{\lambda}^{n} \leq 2^{-n a-\sqrt{n}(R+2 \delta)} \mathbb{1}_{n}\right\}\right) d \mu(\lambda)+2^{-\sqrt{n} \delta} \\
&=1+2^{-\sqrt{n} \delta}-\int_{\mathcal{L}_{=}(a)} \operatorname{Tr}\left(\rho_{\lambda}^{n}\left\{\rho_{\lambda}^{n} \leq 2^{-n a-\sqrt{n}(R+2 \delta)} \mathbb{1}_{n}\right\}\right) d \mu(\lambda) \\
&-\int_{\mathcal{L}_{<}(a)} \operatorname{Tr}\left(\rho_{\lambda}^{n}\left\{\rho_{\lambda}^{n} \leq 2^{-n a-\sqrt{n}(R+2 \delta)} \mathbb{1}_{n}\right\}\right) d \mu(\lambda) \\
&-\int_{\mathcal{L}_{>}(a)} \operatorname{Tr}\left(\rho_{\lambda}^{n}\left\{\rho_{\lambda}^{n} \leq 2^{-n a-\sqrt{n}(R+2 \delta)} \mathbb{1}_{n}\right\}\right) d \mu(\lambda) \tag{5.10}
\end{align*}
$$

where we defined $\mathcal{L}_{>}(a):=\left\{\lambda \in \Lambda: S\left(\rho_{\lambda}\right)>a\right\}$. By Theorem 2.3 and Lemma 5.1 we have the following:

$$
\liminf _{n \rightarrow \infty} \operatorname{Tr}\left(\rho_{\lambda}^{n}\left\{\rho_{\lambda}^{n} \leq 2^{-n a-\sqrt{n}(R+2 \delta)} \mathbb{1}_{n}\right\}\right)= \begin{cases}\Phi\left(\frac{-(R+2 \delta)}{\sigma_{\lambda}}\right) & \text { if } S\left(\rho_{\lambda}\right)=a  \tag{5.11}\\ 1 & \text { if } S\left(\rho_{\lambda}\right)>a \\ 0 & \text { if } S\left(\rho_{\lambda}\right)<a\end{cases}
$$

Taking the limit inferior on both sides of (5.10), noting that we can exchange limit and integral by the Dominated Convergence Theorem, and using (5.11), we obtain

$$
\begin{aligned}
\liminf _{n \rightarrow \infty} \bar{F}\left(\mathfrak{E}_{\text {mix }}^{(n)}, \mathcal{C}_{n}\right) & \leq 1-\int_{\mathcal{L}_{=(a)}} \Phi\left(\frac{-(R+2 \delta)}{\sigma_{\lambda}}\right) d \mu(\lambda)-\int_{\mathcal{L}_{>}(a)} d \mu(\lambda) \\
& =1-\int_{\mathcal{L}_{=}(a)} d \mu(\lambda)+\int_{\mathcal{L}_{=}(a)} \Phi\left(\frac{R+2 \delta}{\sigma_{\lambda}}\right) d \mu(\lambda)-\int_{\mathcal{L}_{>}(a)} d \mu(\lambda) \\
& =\int_{\mathcal{L}_{<}(a)} d \mu(\lambda)+\int_{\mathcal{L}_{=}(a)} \Phi\left(\frac{R+2 \delta}{\sigma_{\lambda}}\right) d \mu(\lambda) \\
& <\int_{\mathcal{L}_{<}(a)} d \mu(\lambda)+\int_{\mathcal{L}_{=}(a)} \Phi\left(\frac{b^{*}}{\sigma_{\lambda}}\right) d \mu(\lambda) \\
& =1-\varepsilon .
\end{aligned}
$$

Here, we used the relation $\Phi(-x)=1-\Phi(x)$ in the first equality, the fact that $\mu$ is a normalized measure on $\Lambda=\mathcal{L}_{=}(a) \cup \mathcal{L}_{<}(a) \cup \mathcal{L}_{>}(a)$ in the second equality, and the assumption $R+2 \delta<b^{*}$ in the strict inequality. This is a contradiction to (5.8a), and hence, we have $b(a, \varepsilon \mid \rho) \geq b^{*}$.

### 5.2.2 Achievability bound

We now use the universal source code $\left\{\mathcal{C}_{n}\right\}_{n \in \mathbb{N}}$ with $\mathcal{C}_{n}:=\left\{\mathcal{V}_{n}, \mathfrak{D}_{n}, M_{n}\right\}$ as defined in Proposition 5.7 to prove that the second order rate $b^{*}$ is achievable. To this end, consider $n$ uses of a mixed source $\left(\rho_{\lambda}, d \mu(\lambda)\right)_{\lambda \in \Lambda}$ with source state $\rho^{(n)}$ as defined in (3.2) and ensemble $\mathfrak{E}_{\text {mix }}^{(n)}$ as defined in (3.3). Recall that $\Pi_{n}$ denotes the projector onto the code space $\Upsilon_{a, b}^{n}$ defined in Section 5.1. For arbitrary $a>0$, the calculation from [7, Sect. V.A.3] shows that we can express the ensemble average fidelity $\bar{F}\left(\mathfrak{E}_{\text {mix }}^{(n)}, \mathcal{C}_{n}\right)$ as

$$
\begin{align*}
\bar{F}\left(\mathfrak{E}_{\text {mix }}^{(n)}, \mathcal{C}_{n}\right) & =\operatorname{Tr}\left(\Pi_{n} \rho^{(n)}\right) \\
& =\int_{\Lambda} \operatorname{Tr}\left(\Pi_{n} \rho_{\lambda}^{\otimes n}\right) d \mu(\lambda) \\
& \geq \int_{\Lambda} \operatorname{Tr}\left[\rho_{\lambda}^{\otimes n}\left\{\rho_{\lambda}^{\otimes n}>\exp (-n a-\sqrt{n} b) \mathbb{1}_{n}\right\}\right] d \mu(\lambda) \\
& =1-\int_{\Lambda} \operatorname{Tr}\left[\rho_{\lambda}^{\otimes n}\left\{\rho_{\lambda}^{\otimes n} \leq \exp (-n a-\sqrt{n} b) \mathbb{1}_{n}\right\}\right] d \mu(\lambda) \tag{5.12}
\end{align*}
$$

where the inequality follows from (5.5). We set $b=b^{*}$, where $b^{*}$ is once again defined as the solution of the relation

$$
\int_{\mathcal{L}_{=(a)}} \Phi\left(\frac{b}{\sigma_{\lambda}}\right) d \mu(\lambda)+\int_{\mathcal{L}_{<}(a)} d \mu(\lambda)=1-\varepsilon
$$

Similar to Section 5.2.1, we then compute

$$
\begin{aligned}
\liminf _{n \rightarrow \infty} \int_{\Lambda} & \operatorname{Tr}\left[\rho_{\lambda}^{\otimes n}\left\{\rho_{\lambda}^{\otimes n} \leq \exp \left(-n a-\sqrt{n} b^{*}\right) \mathbb{1}_{n}\right\}\right] d \mu(\lambda) \\
= & \liminf _{n \rightarrow \infty} \int_{\mathcal{L}_{=}(a)} \operatorname{Tr}\left[\rho_{\lambda}^{\otimes n}\left\{\rho_{\lambda}^{\otimes n} \leq \exp \left(-n a-\sqrt{n} b^{*}\right) \mathbb{1}_{n}\right\}\right] d \mu(\lambda) \\
& \quad+\liminf _{n \rightarrow \infty} \int_{\mathcal{L}_{<}(a)} \operatorname{Tr}\left[\rho_{\lambda}^{\otimes n}\left\{\rho_{\lambda}^{\otimes n} \leq \exp \left(-n a-\sqrt{n} b^{*}\right) \mathbb{1}_{n}\right\}\right] d \mu(\lambda)
\end{aligned}
$$

$$
\begin{aligned}
& +\liminf _{n \rightarrow \infty} \int_{\mathcal{L}_{>}(a)} \operatorname{Tr}\left[\rho_{\lambda}^{\otimes n}\left\{\rho_{\lambda}^{\otimes n} \leq \exp \left(-n a-\sqrt{n} b^{*}\right) \mathbb{1}_{n}\right\}\right] d \mu(\lambda) \\
= & \int_{\mathcal{L}_{=}(a)} \Phi\left(\frac{b^{*}}{\sigma_{\lambda}}\right) d \mu(\lambda)+\int_{\mathcal{L}_{<}(a)} d \mu(\lambda) \\
= & 1-\varepsilon
\end{aligned}
$$

where the exchange of the limit inferior and the integral is permitted by the Dominated Convergence Theorem, and we once again used (5.11). Hence, we obtain $\lim \inf _{n \rightarrow \infty} \bar{F}\left(\mathfrak{E}_{\text {mix }}^{(n)}, \mathcal{C}_{n}\right) \geq 1-\varepsilon$ by (5.12). Moreover, Lemma 5.6 yields that the universal source code $\left\{\mathcal{C}_{n}\right\}_{n \in \mathbb{N}}$ satisfies

$$
\limsup _{n \rightarrow \infty} \frac{\log M_{n}-n a}{\sqrt{n}} \leq b^{*}
$$

Hence, the rate $b^{*}$ is achievable, and we obtain $b(a, \varepsilon \mid \rho) \leq b^{*}$. Together with $b(a, \varepsilon \mid \rho) \geq b^{*}$ from the preceding section, this proves Theorem 4.1.

### 5.3 Mixed source consisting of two memoryless sources

In this section, we prove the second order asymptotic rates for $n$ uses of a mixed source $\left(\rho_{1}, \rho_{2}, t\right)$ consisting of two memoryless sources $\rho_{1}$ and $\rho_{2}$, as stated in Theorem 4.3. We set $S_{i}=S\left(\rho_{i}\right)$ and $\sigma_{i}=\sigma\left(\rho_{i}\right)$ for $i=1,2$. By Corollary 4.2, we have the relation

$$
\begin{equation*}
\sum_{i: S_{i}=a} t_{i} \Phi\left(\frac{L}{\sigma_{i}}\right)+\sum_{i: S_{i}<a} t_{i}=1-\varepsilon \tag{5.13}
\end{equation*}
$$

In the first case of Theorem 4.3 , where $S_{1}=S_{2}=S$, we set $a=S$ in (5.13), which immediately yields

$$
t \Phi\left(\frac{L}{\sigma_{1}}\right)+(1-t) \Phi\left(\frac{L}{\sigma_{2}}\right)=1-\varepsilon
$$

and thus proves Theorem 4.3(i).
Consider now the second case of Theorem 4.3, where $S_{1}>S_{2}$ and $t>\varepsilon$. Choosing $a=S_{1}$, we obtain from (5.13) that

$$
t \Phi\left(\frac{b^{*}}{\sigma_{1}}\right)+1-t=1-\varepsilon
$$

which implies that

$$
b^{*}=\sigma_{1} \Phi^{-1}\left(1-\frac{\varepsilon}{t}\right)=-\sigma_{1} \Phi^{-1}\left(\frac{\varepsilon}{t}\right)
$$

This is the assertion of Theorem 4.3(ii).
Finally, we consider the third case of Theorem 4.3, where $S_{1}>S_{2}$ and $t<\varepsilon$. Choosing $a=S_{2}$ in (5.13) yields

$$
b^{*}=\sigma_{2} \Phi^{-1}\left(\frac{1-\varepsilon}{1-t}\right)=\sigma_{2} \Phi^{-1}\left(1-\frac{\varepsilon-t}{1-t}\right)=-\sigma_{2} \Phi^{-1}\left(\frac{\varepsilon-t}{1-t}\right)
$$

and this proves Theorem 4.3(iii).

## 6 Conclusions and open questions

We derived the second order asymptotic rates of fixed-length visible quantum source coding using a mixed source consisting of memoryless sources. To our knowledge, this is the first example of a second order asymptotic analysis of the optimal rate for a quantum informationprocessing task which uses a resource with memory. Previously, such analyses in the quantum setting were restricted to memoryless (or i.i.d.) resources $[24,15,14,13,7,3]$.

An interesting problem is to extend our methods to mixed classical-quantum channels. In the classical case this has been studied by Polyanskiy et al. [19] (see also [26]). The main result about the second order expansion of the capacity of a mixed channel ([19, Thm. 7$]$ ) bears a close resemblance to the equivalent result about source coding using a mixed source as in [16].

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## References

[1] H. Barnum, C. M. Caves, C. A. Fuchs, R. Jozsa, and B. Schumacher. "On quantum coding for ensembles of mixed states". Journal of Physics A: Mathematical and General 34.35 (2001), p. 6767. arXiv: quant-ph/0008024.
[2] S. Beigi, N. Datta, and F. Leditzky. "Decoding Quantum Information via the Petz recovery map". arXiv preprint (2015). arXiv: 1504.04449 [quant-ph].
[3] S. Beigi and A. Gohari. "Quantum Achievability Proof via Collision Relative Entropy". IEEE Transactions on Information Theory 60.12 (2014), pp. 7980-7986. arXiv: 1312.3822 [quant-ph].
[4] G. Bowen and N. Datta. "Quantum coding theorems for arbitrary sources, channels and entanglement resources". arXiv preprint (2006). arXiv: quant-ph/0610003.
[5] I. Csiszar and J. Körner. Information theory: coding theorems for discrete memoryless systems. Cambridge University Press, 2011.
[6] N. Datta, M.-H. Hsieh, and J. Oppenheim. "An upper bound on the second order asymptotic expansion for the quantum communication cost of state redistribution". arXiv preprint (2014). arXiv: 1409.4352 [quant-ph].
[7] N. Datta and F. Leditzky. "Second-Order Asymptotics for Source Coding, Dense Coding, and Pure-State Entanglement Conversions". IEEE Transactions on Information Theory 61.1 (2015), pp. 582-608. arXiv: 1403.2543 [quant-ph].
[8] N. Datta, M. Tomamichel, and M. M. Wilde. "Second-Order Coding Rates for EntanglementAssisted Communication". arXiv preprint (2014). arXiv: 1405.1797 [quant-ph].
[9] T. S. Han. Information-spectrum methods in information theory. Springer, 2003.
[10] M. Hayashi. "Exponents of quantum fixed-length pure-state source coding". Physical Review A 66.3 (2002), p. 032321. arXiv: quant-ph/0202002.
[11] M. Hayashi. "Second-order asymptotics in fixed-length source coding and intrinsic randomness". IEEE Transactions on Information Theory 54.10 (2008), pp. 4619-4637. arXiv: cs/0503089.
[12] R. Jozsa, M. Horodecki, P. Horodecki, and R. Horodecki. "Universal quantum information compression". Physical Review Letters 81.8 (1998), pp. 1714-1717. arXiv: quant - ph / 9805017.
[13] W. Kumagai and M. Hayashi. "Entanglement Concentration is Irreversible". Physical Review Letters 111.13 (2013), p. 130407. arXiv: 1305.6250 [quant-ph].
[14] W. Kumagai and M. Hayashi. "Second order asymptotics for random number generation". 2013 IEEE International Symposium on Information Theory Proceedings (ISIT). IEEE. 2013, pp. 1506-1510. arXiv: 1303.0669 [cs.IT].
[15] K. Li. "Second order asymptotics for quantum hypothesis testing". Annals of Statistics 42.1 (2014), pp. 171-189. arXiv: 1208.1400 [quant-ph].
[16] R. Nomura and T. S. Han. "Second-order resolvability, intrinsic randomness, and fixedlength source coding for mixed sources: Information spectrum approach". IEEE Transactions on Information Theory 59.1 (2013), pp. 1-16. arXiv: 1106.1879 [cs.IT].
[17] J. R. Norris. Markov chains. Cambridge University Press, 1998.
[18] T. Ogawa and H. Nagaoka. "Strong converse and Stein's lemma in quantum hypothesis testing". IEEE Transactions on Information Theory 46.7 (2000), pp. 2428-2433.
[19] Y. Polyanskiy, H. V. Poor, and S. Verdú. "Dispersion of the Gilbert-Elliott channel". IEEE Transactions on Information Theory 57.4 (2011), pp. 1829-1848.
[20] B. Schumacher. "Quantum coding". Physical Review A 51.4 (1995), p. 2738.
[21] C. E. Shannon. "A mathematical theory of communication". Bell System Technical Journal 27 (1948), pp. 379-423; 623-656.
[22] V. Strassen. "Asymptotische Abschätzungen in Shannons Informationstheorie". Transactions of the Third Prague Conference on Information Theory. 1962, pp. 689-723.
[23] M. Tomamichel, M. Berta, and J. M. Renes. "Quantum Coding with Finite Resources". arXiv preprint (2015). arXiv: 1504.0461 [quant-ph].
[24] M. Tomamichel and M. Hayashi. "A hierarchy of information quantities for finite block length analysis of quantum tasks". IEEE Transactions on Information Theory 59.11 (2013), pp. 7693-7710. arXiv: 1208.1478 [quant-ph].
[25] M. Tomamichel and V. Y. Tan. "Second-Order Asymptotics for the Classical Capacity of Image-Additive Quantum Channels". Communications in Mathematical Physics 338.1 (2015), pp. 103-137. arXiv: 1308.6503 [quant-ph].
[26] M. Tomamichel and V. Y. Tan. "Second-Order Coding Rates for Channels With State". IEEE Transactions on Information Theory 60.8 (2014), pp. 4427-4448. arXiv: 1305.6789 [cs.IT].
[27] H. Umegaki. "Conditional expectation in an operator algebra, IV (entropy and information)". Kodai Mathematical Seminar Reports. Vol. 14. 2. Department of Mathematics, Tokyo Institute of Technology. 1962, pp. 59-85.
[28] A. Winter. "Coding theorem and strong converse for quantum channels". IEEE Transactions on Information Theory 45.7 (1999), pp. 2481-2485. arXiv: 1409.2536 [quant-ph].
[29] A. Winter. "Coding Theorems of Quantum Information Theory". PhD thesis. Universität Bielefeld, 1999. arXiv: quant-ph/9907077.
[30] H. Yagi, T. S. Han, and R. Nomura. "First- and second-order coding theorems for mixed memoryless channels with general mixture". 2015 IEEE International Symposium on Information Theory (ISIT). 2015, pp. 2969-2973. arXiv: 1501.05887 [cs.IT].
[31] H. Yagi and R. Nomura. "Single-letter characterization of epsilon-capacity for mixed memoryless channels". 2014 IEEE International Symposium on Information Theory (ISIT). 2014, pp. 2874-2878. arXiv: 1407.0124 [cs.IT].


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[^1]:    ${ }^{1}$ This approach provides a unifying mathematical framework for obtaining asymptotic rate formulae for various different tasks in information theory, without making any assumptions on the structure or properties of the underlying resources.

[^2]:    ${ }^{2}$ For a discussion of the blind setting and its comparison to the visible setting, see e.g. [1, 10, 29] or Section V.A in [7].

[^3]:    ${ }^{3}$ Note that the assumption $S_{1}>S_{2}$ in Cases 2 and 3 can be made without loss of generality.

