

Part IV

Quantum Information Theory

Chapter 8

Measures of Quantum Information

8.1 Quantum Min-Entropy

We define the quantum counterpart of the classical min-entropy (Definition A.4). By convention, here and throughout the rest of the notes, \log denotes the *binary* logarithm, i.e., the logarithm to base 2.

Definition 8.1. For a given $\rho_A \in \mathcal{D}(A)$ the **min-entropy** of A is

$$H_\infty(A) := H_\infty(\rho_A) := -\log \lambda_{\max}(\rho_A) = -\log \|\rho_A\|_\infty.$$

Obviously, if ρ_X is classical, then the definition coincides with the classical notion. Also, we see that for any $\rho_A \in \mathcal{D}(A)$, its min-entropy is bounded by $0 \leq H_\infty(A) \leq \log \dim(\mathcal{H}_A)$, as for the classical counterpart. However, in certain other aspects, the quantum version behaves very differently. For instance, if AB is an EPR pair, i.e. $\rho_{AB} = |\Phi\rangle\langle\Phi|$ with $|\Phi\rangle = \frac{1}{\sqrt{2}}(|0\rangle|0\rangle + |1\rangle|1\rangle)$, then $\rho_A = \frac{1}{2}|0\rangle\langle 0| + \frac{1}{2}|1\rangle\langle 1|$ and thus

$$H_\infty(A) = 1 > 0 = H_\infty(AB),$$

meaning that the entropy can decrease when considering a larger system. In other words, the quantum version of H_∞ violates **monotonicity** (see Lemma A.4). This is an artifact of entanglement.

How to define the quantum version of the *conditional* min-entropy is way less obvious. We start with the following auxiliary definition.

Definition 8.2. Let $\rho_{AE} \in \mathcal{D}(AE)$ and $\sigma_E \in \mathcal{D}(E)$. Then, the **min-entropy of ρ_{AE} relative to σ_E** is given by

$$H_\infty(\rho_{AE}|\sigma_E) := -\log \min\{\lambda > 0 \mid \lambda \cdot \mathbb{I}_A \otimes \sigma_E \geq \rho_{AE}\}.$$

with the understanding that $H_\infty(\rho_{AE}|\sigma_E) = -\infty$ if no such λ exists.

Remark 8.1. If $\text{supp}(\rho_{AE}) \not\subseteq \text{supp}(\mathbb{I}_A \otimes \sigma_E)$ then there is no λ satisfying the inequality. Indeed, if $|\Omega\rangle \in \ker(\mathbb{I}_A \otimes \sigma_E)$ but not in $\ker(\rho_{AE})$ then $\lambda \cdot \langle\Omega|(\mathbb{I}_A \otimes \sigma_E)|\Omega\rangle = 0 < \langle\Omega|\rho_{AE}|\Omega\rangle$ for any choice of λ (exploiting Remark 0.3). Below, we show that the necessary condition $\text{supp}(\rho_{AE}) \subseteq \text{supp}(\mathbb{I}_A \otimes \sigma_E)$ for $H_\infty(\rho_{AE}|\sigma_E)$ to be finite is equivalent to $\text{supp}(\rho_E) \subseteq \text{supp}(\sigma_E)$ (Corollary 8.2), and that this condition is also sufficient (Lemma 8.3).

Lemma 8.1. Let $0 \leq R_{AE} \in \mathcal{L}(\mathcal{H}_{AE})$, $R_E = \text{tr}_A(R_{AE})$, and $|\varphi\rangle \in \mathcal{H}_E$. Then:

$$|\varphi\rangle \in \ker(R_E) \iff |\psi\rangle|\varphi\rangle \in \ker(R_{AE}) \quad \forall |\psi\rangle \in \mathcal{H}_A.$$

Furthermore, $\ker(\mathbb{I}_A \otimes R_E) = \mathcal{H}_A \otimes \ker(R_E) \subseteq \ker(R_{AE})$.

Proof. Let $\{|i\rangle\}_{i \in I}$ be an arbitrary orthonormal basis of \mathcal{H}_A . The equivalence claim then follows from the observation that

$$\begin{aligned} \sum_i \langle i | \langle \varphi | R_{AE} | i \rangle | \varphi \rangle &= \sum_i \text{tr}(R_{AE}(|i\rangle\langle i| \otimes |\varphi\rangle\langle \varphi|)) \\ &= \text{tr}(R_{AE}(\mathbb{I} \otimes |\varphi\rangle\langle \varphi|)) = \text{tr}(R_E |\varphi\rangle\langle \varphi|) = \langle \varphi | R_E | \varphi \rangle, \end{aligned}$$

together with Remark 0.3 and the positivity of both R_{AE} and R_E .

Regarding the second claim, we first note that the subset-claim follows directly from the proven \Rightarrow -implication. For the equality, consider $|\Phi\rangle \in \mathcal{H}_{AE}$, written as $|\Phi\rangle = \sum_i \alpha_i |i\rangle |\varphi_i\rangle$, and note that

$$\langle \Phi | (\mathbb{I}_A \otimes R_E) | \Phi \rangle = \sum_i |\alpha_i|^2 \langle \varphi_i | R_E | \varphi_i \rangle.$$

Thus, again exploiting Remark 0.3,

$$|\Phi\rangle \in \ker(\mathbb{I}_A \otimes R_E) \iff |\varphi_i\rangle \in \ker(R_E) \forall i \text{ with } \alpha_i \neq 0 \iff |\Phi\rangle \in \mathcal{H}_A \otimes \ker(R_E),$$

which then completes the proof. \square

Corollary 8.2. *For any $0 \leq R_{AE} \in \mathcal{L}(\mathcal{H}_A \otimes \mathcal{H}_E)$ and $0 \leq L_E \in \mathcal{L}(H_E)$:*

$$\text{supp}(R_{AE}) \subseteq \text{supp}(\mathbb{I}_A \otimes L_E) \iff \text{supp}(R_E) \subseteq \text{supp}(L_E).$$

Furthermore, $\text{supp}(\mathbb{I}_A \otimes L_E) = \mathcal{H}_A \otimes \text{supp}(L_E)$.

Proof. We first show “ \Rightarrow ”. Consider $|\varphi\rangle \in \ker(L_E)$ and fix an arbitrary vector $|\psi\rangle \in \mathcal{H}_A$. Then, $(\mathbb{I}_A \otimes L_E)|\psi\rangle|\varphi\rangle = |\psi\rangle \otimes L_E|\varphi\rangle = 0$ and so $|\psi\rangle|\varphi\rangle \in \ker(\mathbb{I}_A \otimes L_E) \subseteq \ker(R_{AE})$, where the inclusion in $\ker(R_{AE})$ is by assumption. By the above lemma, $|\varphi\rangle \in \ker(R_E)$.

For “ \Leftarrow ”, exploiting again the above lemma, we conclude that indeed the assumption implies that $\ker(\mathbb{I}_A \otimes L_E) = \mathcal{H}_A \otimes \ker(L_E) \subseteq \mathcal{H}_A \otimes \ker(R_E) \subseteq \ker(R_{AE})$.

The final claim is obtained from $\mathcal{H}_A \otimes \ker(L_E) = \ker(\mathbb{I}_A \otimes L_E)$, taking orthogonal complements and noting that $(\mathcal{H}_A \otimes \ker(L_E))^\perp = \mathcal{H}_A \otimes \ker(L_E)^\perp$. \square

Lemma 8.3. *For any $\rho_{AE} \in \mathcal{D}(AE)$ and $\sigma_E \in \mathcal{D}(E)$ with $\text{supp}(\rho_E) \subseteq \text{supp}(\sigma_E)$*

$$H_\infty(\rho_{AE} | \sigma_E) = -\log \lambda_{\max}(\sigma_E^{-1/2} \rho_{AE} \sigma_E^{-1/2}) = -\log \|\sigma_E^{-1/2} \rho_{AE} \sigma_E^{-1/2}\|_\infty,$$

where the negative square root of σ_E is by means of its pseudo-inverse.

Proof. Since by assumption and Corollary 8.2, $\text{supp}(\rho_{AE}) \subseteq \text{supp}(\mathbb{I}_A \otimes \sigma_E) = \mathcal{H}_A \otimes \text{supp}(\sigma_E)$, the definition of $H_\infty(\rho_{AE} | \sigma_E)$ as well as the claimed value are not affected when replacing \mathcal{H}_E by $\text{supp}(\sigma_E)$ and considering the corresponding restrictions of ρ_{AE} and σ_E . Therefore, we may assume without loss of generality that σ_E has full rank, and thus is invertible. Then, we see that

$$\lambda \cdot \mathbb{I}_A \otimes \sigma_E \geq \rho_{AE} \iff \lambda \cdot \mathbb{I}_A \otimes \mathbb{I}_E \geq \sigma_E^{-1/2} \rho_{AE} \sigma_E^{-1/2} \iff \lambda \geq \|\sigma_E^{-1/2} \rho_{AE} \sigma_E^{-1/2}\|_\infty,$$

where the second equivalence is easily seen by bringing $\sigma_E^{-1/2} \rho_{AE} \sigma_E^{-1/2}$ into diagonal form. \square

Remark 8.2. If $\lambda \cdot \mathbb{I}_A \otimes \sigma_E \geq \rho_{AE}$ for a $\sigma_E \in \mathcal{D}(E)$ with $\text{supp}(\rho_E) \subsetneq \text{supp}(\sigma_E)$ then we can consider

$$\tilde{\sigma}_E := \frac{\rho_E^0 \sigma_E \rho_E^0}{\text{tr}(\rho_E^0 \sigma_E \rho_E^0)} \in \mathcal{D}(E),$$

which satisfies $\text{supp}(\tilde{\sigma}_E) = \text{supp}(\rho_E)$. Furthermore, using Remark 0.2,

$$\lambda \cdot \text{tr}(\rho_E^0 \sigma_E \rho_E^0) \cdot \mathbb{I}_A \otimes \tilde{\sigma}_E \geq \lambda \cdot \mathbb{I}_A \otimes \rho_E^0 \sigma_E \rho_E^0 \geq \rho_E^0 \rho_{AE} \rho_E^0 = \rho_{AE}.$$

Given that $\text{tr}(\rho_E^0 \sigma_E \rho_E^0) \leq \|\rho_E^0\|_\infty \|\sigma_E\|_1 = 1$ we thus have that $H_\infty(\rho_{AE}|\tilde{\sigma}_E) \geq H_\infty(\rho_{AE}|\sigma_E)$.

Definition 8.3. For any $\rho_{AE} \in \mathcal{D}(AE)$, the **conditional min-entropy** of A given E is defined as

$$H_\infty(A|E) := \sup_{\sigma_E} H_\infty(\rho_{AE}|\sigma_E) = \max_{\sigma_E} H_\infty(\rho_{AE}|\sigma_E)$$

where the supremum/maximum is over all $\sigma_E \in \mathcal{D}(E)$.

Remark 8.3. By Remarks 8.1 and 8.2, the quantification can be restricted to $\sigma_E \in \mathcal{D}(E)$ with $\text{supp}(\sigma_E) = \text{supp}(\rho_E)$, and thus with $\text{supp}(\sigma_E) \subseteq \text{supp}(\rho_E)$. In other words, we may assume without loss of generality that $\mathcal{H}_E = \text{supp}(\rho_E)$. Furthermore, given that the supremum is over a compact set and the objective function is continuous on that set (including the points where the function is $-\infty$), the supremum is attained; thus, writing \max is justified.

In case $E = \emptyset$, i.e., $\mathcal{H}_E = \mathbb{C}$, we obviously have $H_\infty(A|E) = \lambda_{\max}(\rho_A) = H_\infty(A)$. Also, in case of a product state $\rho_{AE} = \rho_A \otimes \rho_E$, we see that $H_\infty(\rho_{AE}|\rho_E) = \lambda_{\max}(\rho_A) = H_\infty(A)$; furthermore, strong subadditivity below implies that $H_\infty(A|E) \leq H_\infty(A)$, and thus we have $H_\infty(A|E) = H_\infty(A)$. For an arbitrary $\rho_{AE} \in \mathcal{D}(\mathcal{H}_A \otimes \mathcal{H}_E)$, we note that strong subadditivity implies that $H_\infty(A|E)$ is still upper bounded by $\log \dim(\mathcal{H}_A)$, and the bound is attained for states of the form $\rho_{AE} = \mu_A \otimes \rho_E = \frac{1}{\dim(\mathcal{H}_A)} \mathbb{I}_A \otimes \rho_E$. On the other hand, looking at the lower bound, it turns out that $H_\infty(A|E)$ may be *negative*!—though not smaller than $-\log \dim(\mathcal{H}_A)$. This is again an artifact of entanglement. For instance, an EPR pair AE has $H_\infty(A|E) = -1$, as we will see later.

Considering the case of a *classical* system conditioned on a quantum system, it is easy to see that if $\rho_{XE} \in \mathcal{D}(\mathcal{X} \otimes \mathcal{H}_E)$ is a hybrid state with classical X then

$$\mathbb{I}_X \otimes \rho_E = \sum_x |x\rangle\langle x| \otimes \rho_E \geq \sum_x |x\rangle\langle x| \otimes P_X(x) \rho_{E|X=x} = \rho_{XE},$$

and thus $H_\infty(X|E) \geq 0$, so this strange behavior does not occur here—because there is no entanglement. Intuitively, if A and E are entangled then ρ_{AE} is not a block-diagonal matrix (and cannot be written as one), like in case of a hybrid state ρ_{XE} above, but it still needs to be “covered” by a block-diagonal matrix, namely by a multiple of $\mathbb{I}_A \otimes \sigma_E$, and thus the latter needs to be “raised higher up”. Note that in the case of a hybrid state ρ_{XE} , we can also write

$$H_\infty(\rho_{XE}|\sigma_E) = -\log \max_x P_X(x) \lambda_{\max}(\sigma_E^{-1/2} \rho_{E|X=x} \sigma_E^{-1/2})$$

The following shows that **monotonicity** is recovered for *classical* subsystems.

Proposition 8.4. Let $\rho_{XAE} \in \mathcal{D}(\mathcal{X} \otimes \mathcal{H}_A \otimes \mathcal{H}_E)$ and $\sigma_E \in \mathcal{D}(\mathcal{H}_E)$. Then

$$H_\infty(\rho_{XAE}|\sigma_E) \geq H_\infty(\rho_{AE}|\sigma_E).$$

Proof. Let $\lambda > 0$ be minimal such that $\lambda \cdot \mathbb{I}_A \otimes \sigma_E \geq \rho_{AE}$, and thus $H_\infty(\rho_{AE}|\sigma_E) = -\log \lambda$. Note that $\rho_{AE} = \sum_x P_X(x) \rho_{AE|X=x}$, and thus $\rho_{AE} \geq P_X(x) \rho_{AE|X=x}$ for all x . It then follows that

$$\lambda \cdot |x\rangle\langle x| \otimes \mathbb{I}_A \otimes \sigma_E \geq P_X(x) |x\rangle\langle x| \otimes \rho_{AE|X=x}.$$

Summing over all x yields that $\lambda \cdot \mathbb{I}_X \otimes \mathbb{I}_A \otimes \sigma_E \geq \rho_{XAE}$, which proves the claim. \square

The following **data-processing inequality** is another natural property: acting on the given system can only make the entropy larger. This in particular implies **strong subadditivity** (as in point 2. of Lemma A.4): $H_\infty(A|BE) \leq H_\infty(A|E)$ for every $\rho_{ABE} \in \mathcal{D}(\mathcal{H}_A \otimes \mathcal{H}_B \otimes \mathcal{H}_E)$.

Proposition 8.5. *Let $\rho_{AE} \in \mathcal{D}(AE)$, $\sigma_E \in \mathcal{D}(E)$ and $\mathfrak{T} \in \mathcal{L}(\mathcal{L}(E), \mathcal{L}(E'))$ a CPTP map. Then*

$$H_\infty(\mathfrak{T}_E(\rho_{AE})|\mathfrak{T}_E(\sigma_E)) \geq H_\infty(\rho_{AE}|\sigma_E).$$

Proof. Let $\lambda > 0$ be minimal such that $\lambda \cdot \mathbb{1}_A \otimes \sigma_E - \rho_{AE} \geq 0$, and thus $H_\infty(\rho_{AE}|\sigma_E) = -\log \lambda$. Because \mathfrak{T} is a CPTP map, also $\lambda \cdot \mathbb{1}_A \otimes \mathfrak{T}_E(\sigma_E) - \mathfrak{T}_E(\rho_{AE}) \geq 0$, which proves the claim. \square

Motivated by the classical definition, let us also here write $\text{Guess}(A|E)$ for $2^{-H_\infty(A|E)}$, keeping in mind though that $\text{Guess}(A|E)$ may not be a probability: it may be greater than 1.

Lemma 8.6. *For any hybrid state $\rho_{YAE} \in \mathcal{D}(\mathcal{Y} \otimes \mathcal{H}_A \otimes \mathcal{H}_E)$ with classical Y we have*

$$\text{Guess}(A|YE) = \sum_y P_Y(y) \text{Guess}(A|E, Y=y),$$

with $\text{Guess}(A|E, Y=y)$ defined by means of the state $\rho_{AE|Y=y}$.

In particular, choosing $E = \emptyset$ and a classical A (referred to as X then), we see that in case of a fully classical state ρ_{XY} , the quantum conditional min-entropy coincides with its classical counterpart. As a matter of fact, for a classical X but a (possibly) quantum E , $\text{Guess}(X|E)$ does coincide with the (optimized) guessing probability of guessing X when given E , i.e.

$$\text{Guess}(X|E) = \sup_{\{M_x\}} \sum_x P_X(x) \text{tr}(M_x^\dagger M_x \rho_E^x)$$

where the supremum is over all measurements $\{M_x\}_{x \in \mathcal{X}}$. The proof is by means of the strong duality property of so-called semidefinite programs; we do not treat this here.

Proof (of Lemma 8.6). By Proposition 8.5, in $\text{Guess}(A|YE) = \min_{\sigma_{YE}} \lambda_{\max}(\sigma_{YE}^{-1/2} \rho_{YAE} \sigma_{YE}^{-1/2})$ it is good enough minimize over all σ_{YE} with classical Y ; indeed, if Y is not classical then we may measure Y , i.e., apply the CPTP map that captures a measurement in the considered basis, which does not affect the state of ρ_{YAE} as there Y is already classical. Thus,

$$\begin{aligned} \text{Guess}(A|YE) &= \min_{Q_Y} \min_{\{\sigma_{E|Y=y}\}} \max_y \frac{P_Y(y)}{Q_Y(y)} \lambda_{\max}(\sigma_{E|Y=y}^{-1/2} \rho_{AE|Y=y} \sigma_{E|Y=y}^{-1/2}) \\ &= \min_{Q_Y} \max_y \frac{P_Y(y)}{Q_Y(y)} \min_{\sigma_{E|Y=y}} \lambda_{\max}(\sigma_{E|Y=y}^{-1/2} \rho_{AE|Y=y} \sigma_{E|Y=y}^{-1/2}) \\ &= \min_{Q_Y} \max_y \frac{P_Y(y)}{Q_Y(y)} \text{Guess}(A|E, Y=y). \end{aligned}$$

We now solve this optimization problem. For this, we observe that the choice of Q_Y for which \max_y is smallest is such that the values \max_y is over are all equal. As such, the minimum is achieved for

$$Q_Y(y) = \frac{P_Y(y) \text{Guess}(A|E, Y=y)}{\sum_{y'} P_Y(y') \text{Guess}(A|E, Y=y')}$$

and it results in the claimed expression. \square

Together with strong subadditivity, this implies the following.