

# Notes on asymptotics of quantum hypothesis testing

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## 1 Preliminaries

Let  $\mathcal{H}$  be a finite dimensional Hilbert space.

### 1.1 Pinching

Let  $A \in B(\mathcal{H})$  be self-djoint, with spectral decomposition  $A = \sum_i \lambda_i P_i$ . We will need the pinching map  $B(\mathcal{H}) \rightarrow B(\mathcal{H})$ , defined as

$$\mathcal{E}_A(X) = \sum_i P_i X P_i.$$

Then  $A$  is a cp unital map. Moreover,  $\mathcal{E}_A(X)$  commutes with  $X$  and we have the pinching inequality [?] ]

$$\mathcal{E}_A(X) \leq |\text{spec}(A)|X, \quad X \geq 0. \quad (1)$$

### 1.2 Relative entropies

Let  $\rho$  and  $\sigma$  be density operators. The (Umegaki) relative entropy is defined as

$$D(\rho\|\sigma) := \begin{cases} \text{Tr} [\rho(\log \rho - \log \sigma)], & \text{supp}(\rho) \leq \text{supp}(\sigma) \\ \infty, & \text{otherwise.} \end{cases}$$

The standard Rényi relative entropy for  $\alpha \in [0, 1] \setminus \{1\}$  is defined as

$$D_\alpha(\rho\|\sigma) := \begin{cases} \frac{1}{\alpha - 1} \log \text{Tr} [\rho^\alpha \sigma^{1-\alpha}], & \text{supp}(\rho) \leq \text{supp}(\sigma) \text{ or } \alpha \in (0, 1) \\ \infty, & \text{otherwise.} \end{cases}$$

The sandwiched Rényi relative entropy for  $\alpha \in [1/2, \infty] \setminus \{1\}$  is defined as

$$\hat{D}_\alpha(\rho\|\sigma) := \begin{cases} \frac{1}{\alpha - 1} \log \text{Tr} [\sigma^{\frac{1-\alpha}{2\alpha}} \rho \sigma^{\frac{1-\alpha}{2\alpha}}], & \text{supp}(\rho) \leq \text{supp}(\sigma) \text{ or } \alpha \in [1/2, 1) \\ \infty, & \text{otherwise.} \end{cases}$$

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### 1.3 The functions $\phi$ and $\psi$

We assume  $\text{supp}(\rho) \leq \text{supp}(\sigma)$ , so that  $D(\rho\|\sigma) < \infty$ .

Let us define

$$\phi(s) = \log \text{Tr} [\rho^{1-s} \sigma^s], \quad s \in \mathbb{R}.$$

Then  $\phi$  is a strictly convex and smooth function, with derivative

$$\phi'(s) = (\text{Tr} [\rho^{1-s} \sigma^s])^{-1} \text{Tr} [\rho^{1-s} \sigma^s (\log \sigma - \log \rho)],$$

[? , Exercise 3.5] In particular,  $\phi'(0) = -D(\rho\|\sigma)$  and  $\phi'(1) = D(\sigma\|\rho)$  Let us define

$$\psi(\lambda) = \min_{s \in \mathbb{R}} \lambda s + \phi(s).$$

**Lemma 1.** *Let  $-D(\sigma\|\rho) = -\phi'(1) \leq \lambda \leq -\phi'(0) = D(\rho\|\sigma)$ . Then*

1.  $\psi(\lambda) = \inf_{s \in [0,1]} \lambda s + \phi(s)$
2.  $\psi$  is monotone increasing
3.  $\psi(-D(\sigma\|\rho)) = -D(\sigma\|\rho)$ ,  $\psi(D(\rho\|\sigma)) = 0$ .

*Proof.* By strict convexity, the derivative  $\phi'(s)$  is increasing, so that there is some  $s_0 \in [0, 1]$  such that  $\lambda = -\phi'(s_0)$  and the function  $s \mapsto \lambda s + \phi(s)$  is decreasing for  $s \leq s_0$  and increasing for  $s_0 \leq s$ . It follows that the infimum is attained at  $s_0 \in [0, 1]$ . This also implies (3).

Assume that  $\lambda \in (-\phi'(1), \phi'(0))$ , then  $\lambda = -\phi'(s_0)$  for some  $s_0 \in (0, 1)$  and we have

$$\psi(\lambda) = \lambda s_0 + \phi(s_0) = \phi(s_0) - \phi'(s_0) s_0 < \phi(0) = 0 = \psi(-\phi(0)),$$

the inequality follows by strict convexity of  $\phi$ . If  $-\phi'(1) \leq \lambda_1 < \lambda$ , then clearly

$$\psi(\lambda_1) \leq \lambda_1 s_0 + \phi(s_0) < \lambda s_0 + \phi(s_0) = \psi(\lambda).$$

This proves (2). □

**Lemma 2.** *Let  $\lambda \in [-D(\sigma\|\rho), D(\rho\|\sigma)]$  and  $0 \leq r \leq D(\rho\|\sigma)$ . Then*

$$\lambda - \psi(\lambda) = r \iff \psi(\lambda) = b(r) := \inf_{s \in [0,1]} \frac{s}{1-s} r + \frac{1}{1-s} \phi(s).$$

*Proof.* Let  $s_\lambda \in [0, 1]$  be such that  $\lambda = -\phi'(s_\lambda)$ , then the assumptions imply that

$$\psi(\lambda) = -\phi'(s_\lambda) s_\lambda + \phi(s_\lambda) = -\phi'(s_\lambda) - r.$$

Solving for  $\phi'(s_\lambda)$ , we get  $-\phi'(s_\lambda) = \frac{1}{1-s_\lambda} (r + \phi(s_\lambda))$ , so that

$$\psi(\lambda) = \frac{s_\lambda}{1-s_\lambda} r + \frac{1}{1-s_\lambda} \phi(s_\lambda).$$

Let

$$g(s) = \frac{s}{1-s} r + \frac{1}{1-s} \phi(s).$$

Then  $g'(s) = \frac{1}{(1-s)^2}(r + \phi(s) + \phi'(s)(1-s))$ . Now note that  $h(s) := \phi(s) + \phi'(s)(1-s)$  satisfies  $h'(s) = \phi''(s)(1-s)$  so that  $h$  is increasing on  $(0, 1)$  (strict) convexity of  $\phi$ , so that  $-D(\rho\|\sigma) = h(0) \leq h(s) \leq h(1) = 0$ . It follows that the derivative  $g'(s)$  changes sign at a unique point  $s_r \in [0, 1]$ , such that  $r + \phi(s_r) + \phi'(s_r)(1-s_r) = 0$ . Comparing this to the above computation, we see that  $s_r = s_\lambda$  and

$$\psi(\lambda) = \min_{s \in [0,1]} g(s) = b(r).$$

□

We define

$$\psi^*(\lambda) = \inf_{t \in [-1,0]} t\lambda + \phi(t).$$

Again, if  $\lambda > D(\rho\|\sigma) = -\phi'(0)$ , then  $t \mapsto t\lambda + \phi(t)$  is strictly increasing at  $t = 0$ , which implies that  $\phi^*(\lambda) < 0$ .

## 1.4 Inequalities

We have two basic inequalities. For  $A, B \geq 0$ , let  $\{A \geq B\}$  be the sum of eigenprojections of  $A - B$  corresponding to nonnegative eigenvalues, similarly  $\{A \leq B\}$ ,  $\{A > B\}$  etc. Then

**Lemma 3** (Quantum Neyman-Pearson). *We have*

$$\min_{0 \leq T \leq I} \text{Tr}[A(I - T)] + \text{Tr}[BT] = \text{Tr}[A\{A \leq B\}] + \text{Tr}[B\{A > B\}].$$

**Lemma 4** (Audenaert et al). *We have for any  $s \in [0, 1]$ ,*

$$\text{Tr}[A\{A \leq B\}] + \text{Tr}[B\{A > B\}] \leq \text{Tr}[A^{1-s}B^s].$$

These statements hold in the von Neumann algebra case as well.

## 1.5 Nussbaum-Szkola probability distributions

Let  $\rho = \sum_i \lambda_i |x_i\rangle\langle x_i|$  and  $\sigma = \sum_j \mu_j |y_j\rangle\langle y_j|$  be the spectral decompositions. The pair  $(P, Q)$  of Nussbaum-Szkola probability distributions related to  $(\rho, \sigma)$  is defined on  $[n] \times [n]$ , here  $n = \dim(\mathcal{H})$  and  $[n] = \{1, \dots, n\}$ . We put

$$P_{ij} = \lambda_i |\langle x_i | y_j \rangle|^2, \quad Q_{ij} = \mu_j |\langle x_i | y_j \rangle|^2.$$

We then have  $D(\rho\|\sigma) = D(P\|Q)$  and  $D_\alpha(\rho\|\sigma) = D_\alpha(P\|Q)$  for all  $\alpha$ .

For  $(\rho^{\otimes n}, \sigma^{\otimes n})$  we get the iid distributions  $(P^n, Q^n)$ . We also have the following result:

**Lemma 5** (Nussbaum-Szkola). *For any test  $T$  and  $c > 0$  we have*

$$\alpha(T) + c\beta(T) \geq \frac{1}{2}(P(\{P \leq cQ\}) + cQ(\{P > cQ\})) = \frac{1}{2} \sum_{ij} \min\{P_{ij}, cQ_{ij}\}$$

*Proof.* Let  $T$  be a projection, then

$$\text{Tr}[\rho T] = \sum_i \lambda_i \langle x_i | T T | x_i \rangle = \sum_{ij} \lambda_i \langle x_i | T | y_j \rangle \langle y_j | T | y_j \rangle = \sum_{ij} \lambda_i |\langle x_i | T | y_j \rangle|^2$$

and similarly for  $\sigma$ . It follows that

$$\alpha(T) + c\beta(T) = \sum_{ij} \lambda_i |\langle x_i | I - T | y_j \rangle|^2 + c \sum_{ij} \mu_j |\langle x_i | T | y_j \rangle|^2.$$

Now we use the inequality

$$a|u - v|^2 + b|v|^2 \geq \frac{1}{2}|u|^2 \min\{a, b\}$$

to lower bound

$$\alpha(T) + c\beta(T) \geq \frac{1}{2} \sum_{ij} |\langle x_i | y_j \rangle|^2 \min\{\lambda_i, c\mu_j\} = \frac{1}{2} \sum_{i,j} \min\{P_{ij}, cQ_{ij}\}$$

By Lemma 3, this inequality holds for all tests. The equality  $\min\{P_{ij}, cQ_{ij}\} = P(\{P \leq cQ\}) + cQ(\{P > cQ\})$  can be easily seen. □

## 2 QHT

Let  $\rho, \sigma$  be a pair of density matrices. We test the hypothesis  $H_0 = \rho$  against the alternative  $H_1 = \sigma$ . A test is given by an operator  $0 \leq T \leq I$ , corresponding to accepting  $H_0$ . The two error probabilities are

$$\alpha(T) = \text{Tr}[(I - T)\rho], \quad \beta(T) = \text{Tr}[T\sigma].$$

We will consider the asymptotic behaviour of the error probabilities

$$\alpha_n(T_n) = \text{Tr}[(I - T_n)\rho_n], \quad \beta_n(T_n) = \text{Tr}[T_n\sigma_n]$$

in testing  $H_0 = \rho_n := \rho^{\otimes n}$  against  $H_1 = \sigma_n := \sigma^{\otimes n}$ .

### 2.1 Quantum Stein's lemma

**We assume  $\text{supp}(\rho) \leq \text{supp}(\sigma)$ , so that  $D(\rho||\sigma) < \infty$ .**

- Quantum Stein's lemma states that if the I. kind error probabilities are constrained as  $\alpha_n(T_n) \leq \epsilon$ , then the II. kind error probabilities go to zero exponentially, with optimal decay rate equal to the relative entropy  $D(\rho||\sigma)$ .
- The strong converse says that if the decay rate of  $\beta_n(T_n)$  is greater than  $D(\rho||\sigma)$ , then  $\alpha_n(T_n) \rightarrow 0$  exponentially.

We will need the following inequalities (3) and (4).

Let  $\lambda \in \mathbb{R}$  and let  $S_n := \{\rho^{\otimes n} > e^{n\lambda} \sigma^{\otimes n}\}$ . Then using Lemma 4 (Audenaert) with  $A = \rho^{\otimes n}$  and  $B = e^{\lambda n} \sigma^{\otimes n}$ , we get for any  $s \in [0, 1]$

$$\alpha_n(S_n) + e^{n\lambda} \beta_n(S_n) \leq \text{Tr} e^{n\lambda s} [(\rho^{\otimes n})^{1-s} (\sigma^{\otimes n})^s] = e^{n\lambda s} (\text{Tr} [\rho^{1-s} \sigma^s])^n = e^{n(\lambda s + \phi(s))}. \quad (2)$$

Hence by taking the infimum over  $s \in [0, 1]$ ,

$$\alpha_n(S_n) \leq e^{n\psi(\lambda)}, \quad \beta_n(S_n) \leq e^{n(-\lambda+\psi(\lambda))} \quad (3)$$

On the other hand, put  $p_n = \text{Tr}[\rho^{\otimes n} S_n]$  and  $q_n = \text{Tr}[\sigma^{\otimes n} S_n]$ . Then  $p_n \geq e^{n\lambda} q_n$  and therefore  $p_n^t \leq e^{n\lambda t} q_n^t$  for any  $t \in [-1, 0]$ . We get

$$\begin{aligned} 1 - \alpha_n(S_n) &= p_n \leq e^{n\lambda t} p_n^{1-t} q_n^t \leq e^{n\lambda t} (p_n^{1-t} q_n^t + (1 - p_n)^{1-t} (1 - q_n)^t) \leq e^{n\lambda t} \text{Tr}[(\rho^{\otimes n})^{1-t} (\sigma^{\otimes n})^t] \\ &= e^{n(\lambda t + \phi(t))} \end{aligned}$$

for all  $t \in [-1, 0]$ . It follows that for any test  $T_n$ , we have

$$\begin{aligned} 1 - \alpha_n(T_n) &= \text{Tr}[\rho^{\otimes n} T_n] = \text{Tr}[(\rho^{\otimes n} - e^{\lambda n} \sigma^{\otimes n}) T_n] + e^{\lambda n} \beta_n(T_n) \leq \text{Tr}[(\rho^{\otimes n} - e^{\lambda n} \sigma^{\otimes n}) S_n] + e^{\lambda n} \beta_n(T_n) \\ &\leq 1 - \alpha_n(S_n) + e^{\lambda n} \beta_n(T_n) \leq e^{n(\lambda t + \phi(t))} + e^{\lambda n} \beta_n(T_n) \end{aligned}$$

and hence

$$\beta_n(T_n) \geq e^{-n\lambda} (1 - \alpha_n(T_n) - e^{n\psi^*(\lambda)}) \quad (4)$$

Let

$$\beta_n(\epsilon) := \min_{0 \leq T_n \leq I} \{\beta_n(T_n) \mid \alpha_n(T_n) \leq \epsilon\}, \quad \epsilon > 0.$$

**Lemma 6** (Quantum Stein's lemma). *For all  $\epsilon \in (0, 1)$ , we have*

$$\lim_{n \rightarrow \infty} -\frac{1}{n} \log \beta_n(\epsilon) = D(\rho \parallel \sigma).$$

*In particular, there exists a sequence  $T_n$  of tests such that  $\alpha_n(T_n) \rightarrow 0$  and  $\lim_n \frac{1}{n} \log \beta_n(T_n) = -D(\rho \parallel \sigma)$ .*

*Proof.* Let  $\lambda < D(\rho \parallel \sigma)$ , then by Lemma 1,  $\psi(\lambda) < 0$ , so that in this case (3),  $\alpha_n(S_n) \rightarrow 0$  and

$$-\frac{1}{n} \log \beta_n(S_n) \geq \lambda - \psi(\lambda) > \lambda.$$

For  $\epsilon \in (0, 1)$  we have  $\alpha_n(S_n) \leq \epsilon$  for large enough  $n$ , so that  $\beta_n(\epsilon) \leq \beta_n(S_n)$ . It follows that

$$\liminf_n -\frac{1}{n} \log \beta_n(\epsilon) \geq \liminf_n -\frac{1}{n} \log \beta_n(S_n) \geq \lambda.$$

Conversely, by (4) we have for any sequence of tests such that  $\alpha_n(T_n) \leq \epsilon$  that

$$\beta_n(T_n) \geq e^{-n\lambda} (1 - \epsilon - e^{n\psi^*(\lambda)}).$$

Since  $\psi^*(\lambda) < 0$  for  $\lambda > D(\rho \parallel \sigma)$ , this implies that, for such  $\lambda$ ,

$$\limsup_n -\frac{1}{n} \log \beta_n(\epsilon) \leq \lambda.$$

Choosing any  $\delta > 0$ , we obtain

$$D(\rho \parallel \sigma) - \delta \leq \liminf_n -\frac{1}{n} \log \beta_n(\epsilon) \leq \limsup_n -\frac{1}{n} \log \beta_n(\epsilon) \leq D(\rho \parallel \sigma) + \delta.$$

Since  $\delta$  was arbitrary, this implies the first statement. For the second statement, we can choose sequences  $\delta_n, \epsilon_n > 0$ ,  $\delta_n, \epsilon_n \rightarrow 0$ , then we can find a sequence of tests  $T_n$  such that  $\alpha_n(T_n) \leq \epsilon_n$  and  $|\frac{1}{n} \log \beta_n(T_n) + D(\rho \parallel \sigma)| < \delta_n$ . □

**Lemma 7** (Strong converse). [?] Let  $T_n$  be a sequence of tests and let  $r > D(\rho\|\sigma)$ . If

$$\limsup_n \frac{1}{n} \log \beta_n(T_n) \leq -r$$

then  $\alpha_n(T_n) \rightarrow 1$  exponentially fast.

*Proof.* By (4), we have for any  $\lambda \in \mathbb{R}$ ,  $\delta > 0$  and large enough  $n$ ,

$$1 - \alpha_n(T_n) \leq e^{n\psi^*(\lambda)} + e^{n\lambda} \beta_n(T_n) \leq e^{n\psi^*(\lambda)} + e^{n(\lambda-r+\delta)}.$$

We then may choose  $\delta$  and  $\lambda$  such that  $D(\rho\|\sigma) < \lambda < r - \delta$ , in which case both of the above exponents are negative. □

## 2.2 Hoeffding bound

The Hoeffding bound studies the exponential decay of  $\alpha_n$  under an exponential constraint on  $\beta_n$ . Specifically, we look at the value of

$$B_e(r) := \sup\left\{-\limsup_n \frac{1}{n} \log \alpha_n(T_n) \mid \limsup_n \frac{1}{n} \log \beta_n(T_n) \leq -r\right\}, \quad r > 0$$

The following result was proved in [?] (direct part) and [?].

**Lemma 8** (Quantum Hoeffding bound). For  $0 < r \leq D(\rho\|\sigma)$ , we have

$$B_e(r) = \sup_{0 \leq s \leq 1} \frac{-sr - \phi(s)}{1-s} = -b(r).$$

*Proof.* By (2), we see that for any  $\lambda$ ,

$$\alpha_n(S_n) \leq e^{n(s\lambda + \phi(s))}, \quad \beta_n(S_n) \leq e^{n(-(1-s)\lambda + \phi(s))}$$

Let us choose  $\lambda$  such that  $-(1-s)\lambda + \phi(s) = -r$ , that is,  $\lambda = \frac{r + \phi(s)}{1-s}$ , then we obtain

$$\alpha_n(S_n) \leq e^{-n \frac{-sr - \phi(s)}{1-s}}, \quad \beta_n(S_n) \leq e^{-nr}.$$

. This implies that  $B_e(r) \geq \sup_{0 \leq s \leq 1} \frac{-sr - \phi(s)}{1-s} = -b(r)$ . To show the lower bound, we will use the pair of Nussbaum-Szkola probability distributions  $(P, Q)$  related to the pair  $(\rho, \sigma)$  and Lemma 5.

What we need to prove is that, for a sequence of tests  $T_n$ ,

$$\limsup_n \frac{1}{n} \log \beta_n(T_n) \leq -r \implies \limsup_n \frac{1}{n} \log \alpha_n(T_n) \geq b(r).$$

By Lemma 5, we have for any  $b \in \mathbb{R}$ ,

$$\alpha_n(T_n) + e^{nb} \beta_n(T_n) \geq \frac{1}{2} [P^n(\{P^n \leq e^{nb} Q^n\}) + e^{-nb} Q^n(\{P^n > e^{nb} Q^n\})] \quad (5)$$

Now note that  $\{P^n \leq e^{nb} Q^n\} = \{\frac{1}{n} \log \frac{Q^n}{P^n} \geq -b\}$  and

$$\log \frac{Q^n(\omega^n)}{P^n(\omega^n)} = \sum_k \log \frac{Q(\omega_k)}{P(\omega_k)}, \quad \omega^n = (\omega_1, \dots, \omega_n) \in \Omega^n,$$

where  $\Omega = [n] \times [n]$ . Put  $X(\omega) = \log \frac{Q(\omega)}{P(\omega)}$ , then  $E_P[X] = -D(P\|Q) = -D(\rho\|\sigma)$  and the cumulant generating function of  $X$  at  $P$  is

$$\log E_P[e^{sX}] = \log E_P[Q^s P^{-s}] = \phi_{P\|Q}(s) = \phi_{\rho\|\sigma}(s) = \phi(s).$$

Now the Cramér theorem of the large deviation theory implies that

$$\lim_n \frac{1}{n} \log(P(\frac{1}{n} \sum_i X_i \geq -b)) = \psi(b)$$

for all  $-b > E_P(X) = -D(\rho\|\sigma)$ . Similarly,  $\{P^n > e^{nb} Q^n\} = \{\frac{1}{n} \sum_i X_i(\omega_i) < -b\}$ , and we have  $E_Q[X] = D(Q\|P) = D(\sigma\|\rho)$  and

$$\log E_Q[e^{sX}] = \log \text{Tr} [\sigma^{1+s} \rho^{-s}] = \phi(1+s).$$

Now note that

$$\inf_s bs + \phi(1+s) = \inf_s b(1+s) + \phi(1+s) - b = \psi(b) - b$$

so that

$$\lim_n \frac{1}{n} \log(e^{nb} Q(\frac{1}{n} \sum_i X_i < -b)) = \psi(b)$$

for  $-b < E_Q[X] = D(\sigma\|\rho)$ . It follows that for any  $b \in (-D(\sigma\|\rho), D(\rho\|\sigma))$ , we have

$$\lim_n \frac{1}{n} \log[P^n(\{P^n \leq e^{nb} Q^n\}) + e^{-nb} Q^n(\{P^n > e^{nb} Q^n\})] = \psi(b),$$

this follows from

$$\log 2 + \min\{\log x, \log y\} = \log(2 \min\{x, y\}) \leq \log(x+y) \leq \log 2 + \max\{\log x, \log y\}$$

and the fact that the two limits above are the same. From (5) and the assumption on  $\beta_n(T_n)$ , we now get

$$\psi(b) \leq \liminf_n \frac{1}{n} \log(\alpha_n(T_n) + e^{nb} \beta_n(T_n)) \leq \max\{\limsup_n \frac{1}{n} \log \alpha_n(T_n), b - r\}.$$

Let us now assume that  $0 < r \leq D(\rho\|\sigma)$  and let  $\lambda \in (-D(\sigma\|\rho), D(\rho\|\sigma)]$  be such that  $r = \lambda - \psi(\lambda)$ . Choose a small  $\epsilon > 0$  such that  $b = \lambda - \epsilon > -D(\sigma\|\rho)$  (we clearly must have  $\lambda > -D(\sigma\|\rho)$ , since otherwise  $r = 0$ ). Then we get

$$\psi(\lambda - \epsilon) \leq \max\{\limsup_n \frac{1}{n} \log \alpha_n(T_n), \psi(\lambda) - \epsilon\}.$$

Since clearly

$$(\lambda - \epsilon)s + \phi(s) > \lambda s + \phi(s) - \epsilon, \quad s \in (0, 1)$$

and by the assumptions both infima are attained in  $(0, 1)$ , we see that  $\phi(\lambda - \epsilon) > \phi(\lambda) - \epsilon$ , so that we must have

$$\phi(\lambda - \epsilon) \leq \limsup_n \frac{1}{n} \log \alpha_n(T_n)$$

Taking the limit  $\epsilon \rightarrow 0$  and noting that  $\phi(\lambda) = b(r)$  finishes the proof. □

### 2.3 Strong converse exponents

By the strong converse of Stein's lemma (Lemma 7), we know that if  $\limsup_n \frac{1}{n} \log \beta_n(T_n) \leq -r$  for  $r > D(\rho\|\sigma)$ , then the success probabilities  $1 - \alpha_n(T_n) \rightarrow 0$  exponentially. We are now interested in the optimal (smallest possible) decay rate, called the strong converse exponent. We put

$$B^*(r) := \inf \left\{ -\liminf_n \frac{1}{n} \log(1 - \alpha_n(T_n)) \mid \limsup_n \frac{1}{n} \log \beta_n(T_n) \leq -r \right\}$$

Let  $\mathcal{M}$  be a von Neumann algebra and let  $\rho, \sigma$  be a pair of normal states. We consider the problem of testing the hypothesis  $H_0 = \rho$  against alternative  $H_1 = \sigma$ . Any test is represented by an operator  $M \in \mathcal{M}$ ,  $0 \leq M \leq I$ , such that the value  $\omega(M)$  is interpreted as the probability of accepting the hypothesis  $H_0$  in the state  $\omega$ .

The two error probabilities related to a test  $M$  are given as

$$\alpha(M) = \rho(1 - M), \quad \beta(M) = \sigma(M).$$

In the Bayes approach, we choose a prior probability  $\pi \in (0, 1)$  and minimize the average error probability over all tests, obtaining the optimal value

$$\begin{aligned} b_\pi(\rho\|\sigma) &:= \inf_M \pi \alpha(M) + (1 - \pi) \beta(M) \\ &= \inf_M \pi \rho(1 - M) + (1 - \pi) \sigma(M) \\ &= \pi - \sup_M (\pi \rho - (1 - \pi) \sigma)(M) = \pi(1 - (\rho - s\sigma)_+(1)) \\ &= \frac{1}{2}(1 - \pi \|\rho - s\sigma\|_1), \quad s = \frac{1 - \pi}{\pi}. \end{aligned}$$

The optimal test for  $\pi \in (0, 1)$  is the quantum Neyman-Pearson test  $M_\pi$ , which is of the form

$$M_\pi = \{\rho > s\sigma\} + X, \quad 0 \leq X \leq \{\rho = s\sigma\}, \quad s = \frac{1 - \pi}{\pi},$$

here for  $\varphi, \psi \in \mathcal{M}_*^+$  we define  $\{\varphi > \psi\}$  as the projection onto the support of  $(\varphi - \psi)_+$ , similarly  $\{\varphi < \psi\}$  is the projection onto the support of  $(\psi - \varphi)_+ = (\varphi - \psi)_-$ . We then have  $\{\varphi \leq \psi\} = 1 - \{\varphi > \psi\}$ ,  $\{\varphi \geq \psi\} = 1 - \{\varphi < \psi\}$  and  $\{\varphi = \psi\} = 1 - \{\varphi < \psi\} - \{\varphi > \psi\}$ .

It is clear from these expressions that the function  $\pi \mapsto b_\pi(\rho\|\sigma)$  is continuous. Furthermore, let us look at the error probabilities for the QNP test  $\{\rho > s\sigma\}$ . Put  $\alpha(s) := \rho(\{\rho \leq s\sigma\})$  and  $\beta(s) := \sigma(\{\rho > s\sigma\})$ . Then By the QNP lemma, we have for any  $s, s' \in (0, \infty)$  that

$$\alpha(s) + s\beta(s) \leq \alpha(s') + s\beta(s') \tag{6}$$

hence

$$s(\beta(s) - \beta(s')) \leq \alpha(s') - \alpha(s) \leq s'(\beta(s) - \beta(s')),$$

the second inequality is obtained by exchanging  $s$  and  $s'$  in (6). It follows that if  $s' > s$ , then  $\beta(s) \geq \beta(s')$  and therefore also  $\alpha(s) \leq \alpha(s')$ .

In the asymmetric QHT approach, we fix the maximum acceptable value  $\epsilon$  of the error  $\alpha(M) \leq \epsilon$  and minimize the error  $\beta(M)$ :

$$d_H^\epsilon(\rho\|\sigma) := \inf \{ \sigma(M), \ 0 \leq M \leq I, \ \rho(1 - M) \leq \epsilon \}.$$

The following result is proved similarly as in the classical case (cf. [? ]).



**Lemma 9.** *We have*

$$b_\pi(\rho\|\sigma) = \inf_{0 < \epsilon < 1} [\pi\epsilon + (1 - \pi)d_H^\epsilon(\rho\|\sigma)], \quad \pi \in (0, 1)$$

*Proof.* Let  $\pi, \epsilon \in (0, 1)$  and let  $M$  be any test such that  $\rho(1 - M) \leq \epsilon$ , then we have

$$b_\pi(\rho\|\sigma) \leq \pi\rho(1 - M) + (1 - \pi)\sigma(M) \leq \pi\epsilon + (1 - \pi)\sigma(M).$$

Taking infimum over all such tests, we obtain

$$b_\pi(\rho\|\sigma) \leq \pi\epsilon + (1 - \pi)d_H^\epsilon(\rho\|\sigma), \quad \forall \epsilon, \pi \in (0, 1). \quad (7)$$

For a fixed  $\pi \in (0, 1)$  and a Neyman-Pearson test  $M_\pi$ , let  $\epsilon_\pi = \rho(1 - M_\pi)$ , then

$$b_\pi(\rho\|\sigma) = \pi\epsilon_\pi + (1 - \pi)\sigma(M_\pi) \geq \pi\epsilon_\pi + (1 - \pi)d_H^{\epsilon_\pi}(\rho\|\sigma),$$

this proves equality. □

## 2.4 In finite dimensions...

**Lemma 10.** *Let  $s(\rho) \leq s(\sigma)$ . Then for any  $\epsilon \in (0, 1)$  there is some  $\pi_\epsilon \in (0, 1)$  and a Neyman-Pearson test  $M_\epsilon = M_{\pi_\epsilon}$  such that  $\rho(1 - M_\epsilon) = \epsilon$ .*

*Proof.* Let

$$s_\epsilon := \sup\{s \geq 0, \text{Tr}[\rho P_{s,-}] < \epsilon\}.$$

Under the assumption, there is some  $\lambda > 0$  such that  $\rho \leq \lambda\sigma$ , and then  $P_{s,-} = I$  for all  $s \geq \lambda$ . It follows that we must have  $c_\epsilon < \lambda$ , in particular,  $c_\epsilon$  is finite. Further, by [?, Lemma], we have for  $s \in [0, \infty)$ :

$$\lim_{t \rightarrow s^-} P_{t,-} = P_{s,-}, \quad \lim_{t \rightarrow s^+} P_{t,-} = P_{s,-} + P_{s,0}$$

and  $P_{s,0} \neq 0$  for a finite number of values of  $s$ . It follows that

$$\text{Tr}[\rho P_{c_\epsilon,-}] = \lim_{t \rightarrow c_\epsilon^-} \text{Tr}[\rho P_{t,-}] \leq \epsilon \leq \lim_{t \rightarrow c_\epsilon^+} \text{Tr}[\rho P_{t,-}] = \text{Tr}[\rho(P_{c_\epsilon,-} + P_{c_\epsilon,0})]$$

and

$$M_\epsilon = P_{c_\epsilon,-} + \frac{\epsilon - \text{Tr}[\rho P_{c_\epsilon,-}]}{\text{Tr}[\rho P_{c_\epsilon,0}]} P_{c_\epsilon,0}$$

is a Neyman-Pearson test for  $\pi_\epsilon := (c_\epsilon + 1)^{-1}$  such that  $\text{Tr}[\rho M_\epsilon] = \epsilon$ . □

**Lemma 11.** *If  $s(\rho) \leq s(\sigma)$  then also*

$$d_H^\epsilon(\rho\|\sigma) = \sup_{0 < \pi < 1} \frac{1}{1 - \pi} [b_\pi(\rho\|\sigma) - \pi\epsilon], \quad \epsilon \in (0, 1).$$

*Proof.* We see that (7) implies

$$d_H^\epsilon(\rho\|\sigma) \geq \frac{1}{1 - \pi} [b_\pi(\rho\|\sigma) - \pi\epsilon]$$

for all  $\epsilon$  and  $\pi$ . If  $s(\rho) \leq s(\sigma)$ , then the second equality follows by Lemma 9. □

### 3 Measured Rényi relative entropy

Let  $\mathcal{M}$  be a von Neumann algebra. A measurement on  $\mathcal{M}$  (with values in a finite set  $\Omega$ ) is defined as a positive linear map  $M : L_\infty(\Omega) \rightarrow \mathcal{M}$  such that  $M(1) = 1$ . Equivalently, a measurement is given by a set of positive operators  $\{a_\omega\}_{\omega \in \Omega}$  such that  $\sum_\omega a_\omega = 1$ . More generally, a measurement can be defined as a unital positive map from a commutative von Neumann algebra to  $\mathcal{M}$ .

Let  $\rho, \sigma$  be normal states on  $\mathcal{M}$ . The measured Rényi relative entropy is defined as

$$D_\alpha^M(\rho \parallel \sigma) = \sup\{D_\alpha(\rho \circ M \parallel \sigma \circ M), M \text{ is a measurement}\}.$$

By [?, Thm. 5.2], we may restrict to finite valued measurements.

It is easily seen that  $D_\alpha^M$  is monotone under positive unital normal maps. Indeed, let  $\Phi : \mathcal{N} \rightarrow \mathcal{M}$  be such a map and let  $M : L_\infty(\Omega) \rightarrow \mathcal{N}$  be a measurement on  $\mathcal{N}$ , then  $\Phi \circ M$  is a measurement on  $\mathcal{M}$  and we have

$$\begin{aligned} D_\alpha^M(\rho \circ \Phi \parallel \sigma \circ \Phi) &= \sup\{D_\alpha(\rho \circ \Phi \circ M \parallel \sigma \circ \Phi \circ M), M \text{ is a measurement on } \mathcal{N}\} \\ &\leq \sup\{D_\alpha(\rho \circ M \parallel \sigma \circ M), M \text{ is a measurement on } \mathcal{M}\} \\ &= D_\alpha^M(\rho \parallel \sigma). \end{aligned}$$

Further, by monotonicity of the sandwiched Rényi relative entropy  $\tilde{D}_\alpha$  for  $\alpha \in [1/2, 1) \cup (1, \infty]$ , and by additivity with respect to tensor products, we have

$$\frac{1}{n} D_\alpha^M(\rho^{\otimes n} \parallel \sigma^{\otimes n}) \leq \frac{1}{n} \tilde{D}_\alpha(\rho^{\otimes n} \parallel \sigma^{\otimes n}) = \tilde{D}_\alpha(\rho \parallel \sigma).$$

We are interested in the equality

$$\lim_n \frac{1}{n} D_\alpha^M(\rho^{\otimes n} \parallel \sigma^{\otimes n}) = \tilde{D}_\alpha(\rho \parallel \sigma), \quad \alpha > 1. \quad (8)$$

This equality was proved in [?] in the finite dimensional case, in [?] in the approximately finite (injective) case and in [?] it was extended to semifinite von Neumann algebras. We will use the Haagerup reduction to show that this equality holds for all von Neumann algebras.

#### 3.1 Haagerup reduction

We just give an outline:

- There exists a von Neumann algebra  $\hat{\mathcal{M}}$  (a crossed product) such that we may identify  $\mathcal{M}$  as a subalgebra in  $\hat{\mathcal{M}}$  and there is a canonical normal conditional expectation  $E_{\mathcal{M}}$  of  $\hat{\mathcal{M}}$  onto  $\mathcal{M}$ ,
- There is an increasing family of subalgebras  $\mathcal{M}_n \subseteq \hat{\mathcal{M}}$  such that
  - Each  $\mathcal{M}_n$  is finite (equipped with a faithful normal tracial state  $\tau_n$ ),
  - $\bigcup_{n \geq 1} \mathcal{M}_n$  is weak\*-dense in  $\hat{\mathcal{M}}$
  - For each  $n$ , there exists a faithful normal conditional expectation  $E_n$  of  $\hat{\mathcal{M}}$  onto  $\mathcal{M}_n$  and for all  $x \in \hat{\mathcal{M}}$ ,  $E_n(x) \rightarrow x$  in the  $\sigma$ -strong topology.

- Any normal state  $\rho$  on  $\mathcal{M}$  extends to a normal state on  $\hat{\mathcal{M}}$  as  $\hat{\rho} := \rho \circ E_{\mathcal{M}}$ . Put  $\rho_n := \hat{\rho}|_{\mathcal{M}_n}$ , then  $\rho_n$  is a normal state on  $\mathcal{M}_n$ . Using the conditional expectation  $E_n$ ,  $\rho_n$  extends to a state  $\hat{\rho}_n = \rho_n \circ E_n = \hat{\rho} \circ E_n$ . We have  $\|\hat{\rho}_n - \hat{\rho}\|_1 \rightarrow 0$ .

We will use this result to prove (8) for any von Neumann algebra  $\mathcal{M}$ . First, note that since  $\rho = \hat{\rho}|_{\mathcal{M}}$  and  $\hat{\rho} = \rho \circ E_{\mathcal{M}}$ , we have by DPI that

$$\tilde{D}_{\alpha}(\rho\|\sigma) = \tilde{D}_{\alpha}(\hat{\rho}\|\hat{\sigma}), \quad D_{\alpha}^M(\rho^{\otimes n}\|\sigma^{\otimes n}) = D_{\alpha}^M(\hat{\rho}^{\otimes n}\|\hat{\sigma}^{\otimes n}).$$

Hence it is enough to work with  $\hat{\rho}$  and  $\hat{\sigma}$ . Notice that by th DPI for  $D_{\alpha}^M$ , we have for any  $m$  and  $\alpha > 1$

$$\lim_n \frac{1}{n} D_{\alpha}^M(\hat{\rho}^{\otimes n}\|\sigma^{\otimes n}) \geq \lim_n \frac{1}{n} D_{\alpha}^M(\hat{\rho}_m^{\otimes n}\|\hat{\sigma}_m^{\otimes n}) = \tilde{D}_{\alpha}(\hat{\rho}_m\|\hat{\sigma}_m),$$

since  $\mathcal{M}_m$  is finite. We therefore have

$$\lim_n \frac{1}{n} D_{\alpha}^M(\hat{\rho}^{\otimes n}\|\hat{\sigma}^{\otimes n}) \geq \lim_m \tilde{D}_{\alpha}(\hat{\rho}_m\|\hat{\sigma}_m) = \tilde{D}_{\alpha}(\hat{\rho}\|\hat{\sigma}).$$

The last equality holds by DPI and lower semicontinuity of  $\tilde{D}_{\alpha}$ . Since the opposite inequality always holds, this proves (8).

## 3.2 Strong converse exponents