

Stein's Lemma in the Quantum Hypothesis Testing

- 1.T. Ogawa and H. Nagaoka, "Strong converse and Stein's lemma in quantum hypothesis testing," in IEEE Transactions on Information Theory, vol. 46, no. 7, pp. 2428-2433, Nov. 2000, doi: 10.1109/18.887855.
- 2.Hiai, F., Petz, D. The proper formula for relative entropy and its asymptotics in quantum probability. *Commun.Math. Phys.* 143, 99–114 (1991). <https://doi.org/10.1007/BF02100287>

Hypothesis Testing

(Recap)

- Let X_1, X_2, \dots, X_n be *i.i.d* $\sim Q(X)$.
- $H_1 : Q = P_1$
- $H_2 : Q = P_2$
- Given $X^n = (X_1, X_2, \dots, X_n)$, We need to decide which hypothesis is true, Or equivalently to say
- $\hat{H} = H_1$ if $g(X^n) = 1$. $\hat{H} = H_2$ if $g(X^n) = 2$.

Error Type

- Type 1 error : $H = H_1, \hat{H} = H_2$ (False alarm)
- Type 2 error: $H = H_2, \hat{H} = H_1$ (Miss detection)
- $\alpha = P(\hat{H} = H_2 | H = H_1)$
- $\beta = P(\hat{H} = H_1 | H = H_2)$

**Question: What is Quantum
version of Hypothesis Testing.**

Quantum Hypothesis testing

- The hypothesis testing problem of two quantum states (Density Operator).

Density Operator

- density matrices, also called density operators, which conceptually take the role of the state vectors, as they encode all the (accessible) information about a quantum mechanical system.

Density Matrix

Properties

- Density operator : $\rho = \sum_i p_i |\rho_i\rangle\langle\rho_i|$
- The expectation value of an observable A in a state, represented by a density matrix ρ , is given by $\langle A \rangle_\rho = \text{tr}(\rho A)$

Quantum Hypothesis Testing

Problem Description

- Let $B(H)$ be the set of linear operators on H .
 $S(H) = \{\rho \in B(H) \mid \rho = \rho^* \geq 0, \text{tr}\{\rho\} = 1\}$.
- Null hypothesis $\rho \in S(H)$ versus alternative hypothesis $\sigma \in S(H)$.
- Decide which hypothesis is true, $\rho^{\otimes n}$ or $\sigma^{\otimes n}$, and the decision is given by a two-valued quantum measurement $\{A_n, I - A_n\}$ ($A_n \in B(H^{\otimes n}), 0 \leq A_n \leq I$)
- A_n corresponds acceptance of $\rho^{\otimes n}$ and $I - A_n$ corresponds acceptance of $\sigma^{\otimes n}$.

Types of Errors

$$\alpha_n(A_n) = \text{tr}(\rho^{\otimes n}(I - A_n))$$

$$\beta_n(A_n) = \text{tr}(\sigma^{\otimes n}A_n)$$

$\alpha_n(A_n)$ is the error probability of the acceptance of $\sigma^{\otimes n}$ when $\rho^{\otimes n}$ is true. (Type 1)

$\beta_n(A_n)$ is the error probability of the converse situation. (Type 2)

Asymmetric v.s. Symmetric case

- Asymmetric case :

$$\beta_n^*(\epsilon) = \min\{\beta_n(A_n) \mid A_n \in B(H^{\otimes n}), 0 \leq A_n \leq I, \alpha_n(A_n) \leq \epsilon\}$$

- Symmetric case:

$$p_{err}^*(p, q) = \min\{p \cdot \alpha_n(A_n) + q \cdot \beta_n(A_n) \mid A_n \in B(H^{\otimes n}), 0 \leq A_n \leq I, p + q = 1\}$$

**Weak Converse property
versus**

Strong Converse property

Asymmetric Quantum Hypothesis Testing

Weak converse property

$$\lim_{n \rightarrow \infty} \sup \frac{1}{n} \log \beta_n^*(\epsilon) \leq -D(\rho || \sigma) \quad (1)$$

$$-\frac{1}{1 - \epsilon} D(\rho || \sigma) \leq \lim_{n \rightarrow \infty} \inf \frac{1}{n} \log \beta_n^*(\epsilon) \quad (2)$$

$$D(\rho || \sigma) = \text{tr}(\rho(\log \rho - \log \sigma))$$

Quantum Data Processing Inequality(QDPI)

For any quantum channel $\mathcal{E} : \mathcal{H}_A \rightarrow \mathcal{H}_B$, and density operator $\rho, \sigma \in \mathcal{D}(\mathcal{H}_A)$:

$$D(\mathcal{E}(\rho) || \mathcal{E}(\sigma)) \leq D(\rho || \sigma)$$

Proof

$$D(\rho^{\otimes n} || \sigma^{\otimes n})$$

$$\geq \alpha_n(A_n) \log \frac{\alpha_n(A_n)}{1 - \beta_n(A_n)} + (1 - \alpha_n(A_n)) \log \frac{1 - \alpha_n(A_n)}{\beta_n(A_n)} \dots \text{(QDPI)}$$

$$\geq \alpha_n(A_n)(\log \alpha_n(A_n) - \log(1 - \beta_n(A_n))) + (1 - \alpha_n(A_n))(\log(1 - \alpha_n(A_n)) - \log \beta_n(A_n))$$

$$\geq \alpha_n(A_n)(\log(\alpha_n(A_n))) + (1 - \alpha_n(A_n))(\log(1 - \alpha_n(A_n))) - \alpha_n(A_n) \log(1 - \beta_n(A_n)) - (1 - \alpha_n(A_n)) \log \beta_n(A_n)$$

$$\geq -H(\alpha_n(A_n)) - \alpha_n(A_n) \log(1 - \beta_n(A_n)) - (1 - \alpha_n(A_n)) \log \beta_n(A_n)$$

$$\geq -\log 2 - (1 - \alpha_n(A_n)) \log \beta_n(A_n) \dots (1)$$

Continue

From equation (1):

We have :

$$(1 - \alpha_n(A_n)) \frac{1}{n} \log \beta_n(A_n) \geq -\frac{\log 2}{n} - D(\rho || \sigma) \quad \dots (2)$$

From equation (2) and $(1 - \alpha_n(A_n) \leq 1 - \epsilon)$ we have:

$$(1 - \epsilon) \frac{1}{n} \log \beta_n(A_n) \geq -\frac{\log 2}{n} - D(\rho || \sigma) \quad \dots (3)$$

From equation (3) when $n \rightarrow \infty$:

$$-\frac{1}{1 - \epsilon} D(\rho || \sigma) \leq \liminf_{n \rightarrow \infty} \frac{1}{n} \log \beta_n^*(\epsilon)$$

Asymmetric Quantum Hypothesis Testing

Weak converse property

$$(1 - \alpha_n(A_n)) \frac{1}{n} \log \beta_n(A_n) \geq -\frac{\log 2}{n} - D(\rho || \sigma) \quad \dots (2)$$

From equation (2), setting $\beta_n(A_n) \leq e^{-nr}$ for $r > D(\rho || \sigma)$. Then we will get:

$$1 - \alpha_n(A_n) \leq \frac{-\log 2 - nD(\rho || \sigma)}{\log \beta_n}$$

$$\Rightarrow \alpha_n(A_n) - 1 \geq \frac{\log 2 + nD(\rho || \sigma)}{\log \beta_n} \geq \frac{\log 2 + nD(\rho || \sigma)}{-nr}$$

$$\Rightarrow \alpha_n(A_n) \geq \frac{nD(\rho || \sigma)}{-nr} + 1 = \frac{D(\rho || \sigma)}{-r} + 1 > 0 \quad (\text{Since } r > D(\rho || \sigma))$$

Asymmetric Quantum Hypothesis Testing

Weak converse property

Theorem: if $\beta_n(A_n) \leq e^{-nr}$ ($r > D(\rho || \sigma)$), then $\alpha_n(A_n)$ does not go to zero as $n \rightarrow \infty$.

Asymmetric Quantum Hypothesis Testing

Strong Converse property

What we want to show:

if $\beta_n(A_n) \leq e^{-nr}$ ($r > D(\rho || \sigma)$), then $\alpha_n(A_n)$ goes to one as $n \rightarrow \infty$.

Asymmetric Quantum Hypothesis Testing

Strong Converse property

Lemma 1: For any test A_n , we have :

$$\text{tr}(\rho^{\otimes n} - e^{n\lambda} \sigma^{\otimes n} X_{n,\lambda}) \geq \text{tr}(\rho^{\otimes n} - e^{n\lambda} \sigma^{\otimes n} A_n)$$

Theorem 1: For any test A_n and any $\lambda \in R$, we have:

$$1 - \alpha_n(A_n) \leq e^{-n\varphi(\lambda)} + e^{n\lambda} \beta_n(A_n)$$

Eigen(spectral)-decomposition:

$$\rho^{\otimes n} - e^{n\lambda} \sigma^{\otimes n} = \sum_j \mu_{n,j} E_{n,j}$$

Where

$$\varphi(\lambda) = \max_{0 \leq s \leq 1} \{ \lambda s - \psi(s) \}$$

$$X_{n,\lambda} = \sum_{j \in D_n} E_{n,j} \quad \text{where } D_n = \{j \mid \mu_{n,j} > 0\}$$

$$\psi(s) = \log \text{tr}(\rho^{1+s} \sigma^{-s})$$

Lemma 1 proof

$$\text{tr}((\rho^{\otimes n} - e^{n\lambda}\sigma^{\otimes n})A_n) = \sum_j \mu_{n,j} \text{tr}(E_{n,j}A_n)$$

$$\leq \sum_{j \in D_n} \mu_{n,j} \text{tr}(E_{n,j}A_n)$$

$$\leq \sum_{j \in D_n} \mu_{n,j} \text{tr}(E_{n,j})$$

$$\text{tr}((\rho^{\otimes n} - e^{n\lambda}\sigma^{\otimes n})X_{n,j})$$

Convexity of $\psi(s)$

Let $u = \rho^{1+s}\sigma^{-s}$, Then $\psi(s) = \log \operatorname{tr}(u)$

$$\psi'(s) = \frac{1}{\operatorname{tr}(u)} \frac{d}{ds}(\operatorname{tr}(u))$$

$$\frac{d}{ds} \operatorname{tr}(u) = \frac{d}{ds}(\operatorname{tr}(\rho^{1+s}\sigma^{-s}))$$

$$= \operatorname{tr}\left(\frac{d}{ds}(\rho^{1+s}\sigma^{-s})\right)$$

$$= \operatorname{tr}(\rho^{1+s}\sigma^{-s}(\log \rho - \log \sigma))$$

$$\varphi(\lambda) = \max_{0 \leq s \leq 1} \{\lambda s - \psi(s)\}$$

$$\psi(s) = \log \operatorname{tr}(\rho^{1+s}\sigma^{-s})$$

Continue the proof

Combine all of them we got:

$$\psi'(s) = e^{-\psi(s)} \text{tr}(\rho^{1+s} \sigma^{-s} (\log \rho - \log \sigma))$$

$$\text{Let } A = \log \rho - \log \sigma - \psi'(s)$$

$$\begin{aligned} \psi''(s) &= e^{-\psi(s)} \text{tr}(\rho^{1+s} A \sigma^{-s} A) \\ &= e^{-\psi(s)} \text{tr}((\rho^{\frac{1+s}{2}} A \sigma^{-\frac{s}{2}})(\rho^{\frac{1+s}{2}} A \sigma^{-\frac{s}{2}})^*) \\ &> 0 \quad \dots\dots (a) \end{aligned}$$

Observation

More observation:

$$(1) \psi(0) = 0$$

$$(2) \psi'(0) = D(\rho || \sigma)$$

$$(3) \varphi(\lambda) > 0 \text{ if } \lambda > D(\rho || \sigma)$$

$$(4) s^* = \arg \max_{0 \leq s \leq 1} \{ \lambda s - \psi(s) \} \iff \psi'(s^*) = \lambda \text{ if } D(\rho || \sigma) \leq \lambda \leq \psi'(1)$$

Proof of Theorem 1

We define two probability distribution: $p_n = \{p_{n,j}\}$, $q = \{q_{n,j}\}$

$p_{n,j} = \text{tr}(\rho^{\otimes n} E_{n,j})$, $q_{n,j} = \text{tr}(\sigma^{\otimes n} E_{n,j})$ (Since $E_{n,j}$ are eigen-decomposition they are orthogonal to each other (i.e. $E_{n,j} E_{n,k} = 0$ if $j \neq k$))

$$\mu_{n,j} \text{tr}(E_{n,j}) = p_{n,j} - e^{n\lambda} q_{n,j}$$

$$\mu_{n,j} \geq 0 \iff p_{n,j} - e^{n\lambda} q_{n,j} \geq 0$$

$$D_n = \{j \mid 0 \leq \forall s \leq 1, e^{-n\lambda s} p_{n,j}^s q_{n,j}^{-s} \geq 1\}$$

Continue

A function f is said to be matrix convex of order n if for all $n \times n$ Hermitian matrices A and B and for all real numbers $0 \leq \lambda \leq 1$:

$$f((1 - \lambda)A + \lambda B) \leq (1 - \lambda)f(A) + \lambda f(B)$$

$$\text{tr}(\rho^{\otimes n} X_{n,j}) = \sum_{j \in D_n} \text{tr}(\rho^{\otimes n} E_{n,j})$$

$$= \sum_{j \in D_n} p_{n,j}$$

$$\leq \sum_{j \in D_n} p_{n,j} \cdot e^{-n\lambda s} p_{n,j}^s q_{n,j}^{-s}$$

$$\leq e^{-n\lambda s} \sum_j p_{n,j}^{1+s} q_{n,j}^{-s} \quad (\text{convex of } q)$$

$$\leq e^{-n\lambda s} \text{tr}((\rho^{\otimes n})^{1+s} (\sigma^{\otimes n})^{-s}) \dots (4)$$

Note:

$f(u) = u^{-s} (0 \leq s \leq 1)$ is a convex function

Detail

$$\begin{aligned} \text{tr}((\rho^{\otimes n})^{1+s}(\sigma^{\otimes n})^{-s}) &= \text{tr}\left(\sum_j E_{n,j}(\rho^{\otimes n})^{1+s}E_{n,j}(\sigma^{\otimes n})^{-s}\right) \\ &\geq \sum_j \text{tr}(E_{n,j}(\rho^{\otimes n})^{1+s})\text{tr}(E_{n,j}(\sigma^{\otimes n})^{-s}) \end{aligned}$$

Continue

We have :

$$tr(\rho^{\otimes n} X_{n,j}) \leq e^{-n(\lambda s - \psi(s))}$$

Hence we know that :

$$tr(\rho^{\otimes n} X_{n,j}) \leq e^{-n\varphi(\lambda)} \text{ by taking the maximum}$$

Finally we have:

$$\begin{aligned} 1 - \alpha_n(A_n) &= tr(\rho^{\otimes n} A_n) \\ &\leq tr(\rho^{\otimes n} - e^{n\lambda} \sigma^{\otimes n}) X_{n,j} + e^{n\lambda} tr(\sigma^{\otimes n} A_n) \\ &\leq tr(\rho^{\otimes n} X_{n,j}) + e^{n\lambda} tr(\sigma^{\otimes n} A_n) \\ &\leq e^{-n\varphi(\lambda)} + e^{n\lambda} \beta_n(A_n) \quad \text{Proved!} \end{aligned}$$

Quantum Stein's lemma

Theorem 2: For any $0 \leq \epsilon < 1$ it holds that :

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

Proof of theorem 2

From theorem 1 we know that :

$$1 - \epsilon \leq 1 - \alpha_n(A_n) \leq e^{-n\varphi(\lambda)} + e^{n\lambda}\beta_n(A_n)$$

$$\implies \beta_n(A_n) \geq e^{-n\lambda}(1 - \epsilon - e^{-n\varphi(\lambda)})$$

Let $\lambda = D(\rho || \sigma + \delta)$ ($\delta > 0$), From property of $\varphi(\lambda)$ we know that $\varphi(\lambda) > 0$ in this case

Hence $1 - \epsilon - e^{-n\varphi(\lambda)} > 0$ for n sufficiently large.

This implies : $\frac{1}{n} \log \beta_n^*(\epsilon) \geq -\lambda + \frac{1}{n} \log(1 - \epsilon - e^{-n\varphi(\lambda)})$

$$\implies \liminf_{n \rightarrow \infty} \geq -D(\rho || \sigma) - \delta \text{ for } \forall \delta > 0$$

Strong Converse

Theorem 3: For any test A_n , if

$$\limsup_{n \rightarrow \infty} \frac{1}{n} \log \beta_n(A_n) \leq -r, \dots\dots(13)$$

Then

$$\limsup_{n \rightarrow \infty} \frac{1}{n} \log(1 - \alpha_n(A_n)) \leq -\varphi(\lambda^*) \dots\dots (14)$$

Where λ^* is a real number which satisfies $\varphi(\lambda^*) = r - \lambda^*$. Moreover, $\varphi(\lambda^*)$ is a represented as:

$$\varphi(\lambda^*) = \max_{0 \leq s \leq 1} \left\{ \frac{s}{1+s} r - \frac{1}{1+s} \psi(s) \right\}. \dots\dots (15)$$

Proof of (14)

For all $\delta > 0$, there exists n_0 such that:

$$\beta_n(A_n) \leq e^{-n(r-\delta)}, \forall n \geq n_0 \text{ from (13).}$$

Put $\lambda = \lambda^*$ in theorem 1, we have:

$$1 - \alpha_n(A_n) \leq e^{-n\varphi(\lambda^*)} + e^{-n(r-\lambda^*-\delta)}, \forall n \geq n_0$$

$$\iff 1 - \alpha_n(A_n) \leq 2e^{-n(\varphi(\lambda^*)-\delta)}$$

$$\iff \limsup_{n \rightarrow \infty} \frac{1}{n} \log(1 - \alpha_n(A_n)) \leq -\varphi(\lambda^*) + \delta$$

Since δ is arbitrary, (14) has been proved!

Proof of (15)

Suppose $\psi'(0) \leq r \leq 2\psi'(1) - \psi(1)$,

Define: $u(r) = \varphi(\lambda^*) = \max_{0 \leq s \leq 1} \{s\lambda^* - \psi(s)\} = r - \lambda^*$

Using observation (4) $\lambda^* = \psi'(s^*)$ we mentioned before:

$u(r) = s^*\psi'(s^*) - \psi(s^*)$ where $r = (s^* + 1)\psi'(s^*) - \psi(s^*)$

Combine both we got $u(r) = \frac{s^*}{s^* + 1}r - \frac{1}{s^* + 1}\psi(s^*)$.

Continue

Let's see the derivative of function:

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

$$g'(s) = \frac{1}{(s+1)^2}(r + \psi(s) - (1+s)\psi'(s))$$

We just need to see how function $h(s) = r + \psi(s) - (1+s)\psi'(s)$ works.

$h'(s) = -(1+s)\psi''(s) \leq 0$ by equation (a), which says that the sign of

$g'(s)$ changes at most once.

Therefore $g(s)$ get maximum value at $h(s) = 0 \iff r = (s+1)\psi'(s) - \psi(s)$

So we got $u(r) = \max_{0 \leq s \leq 1} g(s)$

Clearly if $r \geq 2\psi'(1) - \psi(1)$:

$$\varphi(\lambda^*) = \frac{1}{2}r - \frac{1}{2}\psi(1) = g(1) = \max_{0 \leq s \leq 1} g(s)$$

If $r \leq \psi'(0)$:

$$\varphi(\lambda^*) = 0 = g(0) = \max_{0 \leq s \leq 1} g(s)$$

Strong Converse

Corollary 1: For any test A_n , if

$$\limsup_{n \rightarrow \infty} \frac{1}{n} \log \beta_n(A_n) < -D(\rho || \sigma)$$

Then $\alpha_n(A_n)$ goes to one exponentially .

Proof: set $r = -D(\rho || \sigma) - \delta, \forall \delta > 0$.

We will get $1 - \alpha_n(A_n) \leq 2^{-n(D(\rho || \sigma) - \delta - \lambda^*)}$ by theorem 3.