

On Robertson-type uncertainty principles

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Basic Notations in Quantum Mechanics

State space: the set of $n \times n$ positive definite trace one matrices (\mathcal{M}_n^1).

Observables: $n \times n$ self adjoint matrices ($M_{n,sa}$).

For given state $D \in \mathcal{M}_n^1$ and observables $A, B \in M_{n,sa}$

expectation value: $\text{Tr}(DA)$;

normalization of A: $A_0 = A - \text{Tr}(DA)I$; ($\text{Tr}(DA_0) = 0$)

variance: $\text{Var}_D(A) = \text{Tr}(DA^2) - (\text{Tr}(DA))^2$;

covariance: $\text{Cov}_D(A, B) = \frac{1}{2}(\text{Tr}(DAB) + \text{Tr}(DBA)) - \text{Tr}(DA)\text{Tr}(DB)$.

Uncertainty Relations in Early Years

1927, Heisenberg: Defined the uncertainty of a Gaussian distribution f as its width D_f . The width of the Fourier transformation of f is denoted by $D_{\mathcal{F}(f)}$. The first formalization of the uncertainty principle was $D_f D_{\mathcal{F}(f)} = \text{constant}$.

1927, Kennard: Observables A, B with $[A, B] = -i$: $\text{Var}_D(A) \text{Var}_D(B) \geq \frac{1}{4}$.

1929, Robertson: For every observables A, B and state D : $\text{Var}_D(A) \text{Var}_D(B) \geq \frac{1}{4} |\text{Tr}(D[A, B])|^2$.

1930, Schrödinger: $\text{Var}_D(A) \text{Var}_D(B) - \text{Cov}_D(A, B)^2 \geq \frac{1}{4} |\text{Tr}(D[A, B])|^2$.

1934, Robertson: For every set of observables $(A_i)_{i=1, \dots, N}$

$$\det\left([\text{Cov}_D(A_h, A_j)]_{h,j=1, \dots, N}\right) \geq \det\left(\left[-\frac{i}{2} \text{Tr}(D[A_h, A_j])\right]_{h,j=1, \dots, N}\right).$$

(For $N = 2$ gives the Schrödinger uncertainty relation.)

Quantum Fisher Information

\mathcal{F}_{op} : set of operator monotone functions $f: \mathbb{R}^+ \rightarrow \mathbb{R}$ with properties $f(x) = xf(x^{-1})$ and $f(1) = 1$.

Examples in \mathcal{F}_{op} : $f_{\text{RLD}}(x) = \frac{2x}{1+x}$, $f_{\text{SLD}}(x) = \frac{1+x}{2}$, $f_{\text{WY}}(x) = \left(\frac{1+\sqrt{x}}{2}\right)^2$, $f_{\text{KM}}(x) = \frac{x-1}{\log x}$.

Regular and non-regular elements: $\mathcal{F}_{\text{op}}^r = \{f \in \mathcal{F}_{\text{op}} | f(0) \neq 0\}$ and $\mathcal{F}_{\text{op}}^n = \{f \in \mathcal{F}_{\text{op}} | f(0) = 0\}$.

Theorem [Gibilisco, Hansen, Isola]. The map

$$\mathcal{F}_{\text{op}}^r \rightarrow \mathcal{F}_{\text{op}}^n \quad f \mapsto \tilde{f}(x) = \frac{1}{2} \left[(1+x) - (1-x)^2 \frac{f(0)}{f(x)} \right]$$

is a bijection.

For every $f \in \mathcal{F}_{\text{op}}$ introduce the notation $m_f: \mathbb{R}^+ \times \mathbb{R}^+ \rightarrow \mathbb{R}^+$, $m_f = yf\left(\frac{x}{y}\right)$.

(The reciprocal of m_f is the Chentsov–Morozova function.)

Theorem [Petz]. In quantum setting there is a bijective correspondence between Fisher informations and functions in $f \in \mathcal{F}_{\text{op}}$. For every $f \in \mathcal{F}_{\text{op}}$ the Fisher information is given by

$$\langle A, B \rangle_{D,f} = \text{Tr}\left(A m_f(L_D, R_D)^{-1}(B)\right), \quad (1)$$

where $L_D(X) = DX$, $R_D(X) = XD$.

\Rightarrow For every $f \in \mathcal{F}_{\text{op}}$ ($\mathcal{M}_n, \langle \cdot, \cdot \rangle_{D,f}$) is a Riemannian manifold.

Covariances

For two observables $A, B \in M_{n,sa}$, state $D \in \mathcal{M}_n^1$ and function $f \in \mathcal{F}_{\text{op}}$ we define **covariance:**

$$\text{Cov}_D(A, B) = \frac{1}{2}(\text{Tr}(DAB) + \text{Tr}(DBA)) - \text{Tr}(DA)\text{Tr}(DB);$$

quantum f -covariance: [introduced by Petz]

$$\text{Cov}_D^f(A, B) = \text{Tr}\left(Af(L_{n,D}R_{n,D}^{-1})R_{n,D}(B)\right);$$

antisymmetric f -covariance:

$$\text{qCov}_{D,f}^{as}(A, B) = \frac{f(0)}{2} \langle i[D, A], i[D, B] \rangle_{D,f};$$

symmetric f -covariance:

$$\text{qCov}_{D,f}^s(A, B) = \frac{f(0)}{2} \langle \{D, A\}, \{D, B\} \rangle_{D,f},$$

where $[\cdot, \cdot]$ is the commutator of matrices and $\{\cdot, \cdot\}$ denotes the anticommutator respectively.

For a fixed density matrix $D \in \mathcal{M}_n^1$, function $f \in \mathcal{F}_{\text{op}}$ and an N -tuple of nonzero matrices $(A^{(k)})_{k=1, \dots, N} \in M_{n,sa}$ we define the following $N \times N$ matrices Cov_D , Cov_D^f , $\text{qCov}_{D,f}^{as}$ and $\text{qCov}_{D,f}^s$ with entries

$$\begin{aligned} [\text{Cov}_D]_{ij} &= \text{Cov}_D(A_0^{(i)}, A_0^{(j)}) & [\text{Cov}_D^f]_{ij} &= \text{Cov}_D^f(A_0^{(i)}, A_0^{(j)}) \\ [\text{qCov}_{D,f}^{as}]_{ij} &= \text{qCov}_{D,f}^{as}(A_0^{(i)}, A_0^{(j)}) & [\text{qCov}_{D,f}^s]_{ij} &= \text{qCov}_{D,f}^s(A_0^{(i)}, A_0^{(j)}) \end{aligned}$$

Mathematical Toolbox

Petz's scalar product (1) can be extended: Define

$$\mathcal{C}_{\mathcal{M}} = \left\{ g: \mathbb{R}^+ \times \mathbb{R}^+ \rightarrow \mathbb{R}^+ \mid \begin{array}{l} g \text{ is a symmetric smooth function, with analytical} \\ \text{extension defined on a neighborhood of } \mathbb{R}^+ \times \mathbb{R}^+ \end{array} \right\}.$$

Fix a function $g \in \mathcal{C}_{\mathcal{M}}$. Define for every $D \in \mathcal{M}_n$ and for every $A, B \in M_{n,sa}$

$$\langle A, B \rangle_{D,g} = \text{Tr}\left(Ag(L_{n,D}, R_{n,D})(B)\right).$$

\Rightarrow For every $g \in \mathcal{C}_{\mathcal{M}}$ ($\mathcal{M}_n, \langle \cdot, \cdot \rangle_{D,g}$) is a Riemannian manifold.

Connection to Petz's scalar product: For $f \in \mathcal{F}_{\text{op}}$ define $g(x, y) = \frac{1}{yf\left(\frac{x}{y}\right)}$, then we have

$$\langle A, B \rangle_{D,f} = \langle A, B \rangle_{D,g} \quad \forall D \in \mathcal{M}_n, \forall A, B \in M_{n,sa}.$$

Theorem [Andai, Lovas]. Consider a density matrix $D \in \mathcal{M}_n^1$, an N -tuple of observables $(A^{(k)})_{k=1, \dots, N}$ and functions $g_1, g_2 \in \mathcal{C}_{\mathcal{M}}$ such that

$$g_1(x, y) \geq g_2(x, y) \quad \forall x, y \in \mathbb{R}^+.$$

Define the $N \times N$ matrices \mathbf{Cov}_{D,g_1} and \mathbf{Cov}_{D,g_2} with entries $[\mathbf{Cov}_{D,g_k}]_{ij} = \langle A_0^{(i)}, A_0^{(j)} \rangle_{D,g_k}$ ($k = 1, 2$). Then

$$\det(\mathbf{Cov}_{D,g_1}) \geq \det(\mathbf{Cov}_{D,g_2}) + \det(\mathbf{Cov}_{D,g_1} - \mathbf{Cov}_{D,g_2})$$

holds.

Uncertainty Relations Nowadays

Gibilisco and Isola in 2006 conjectured the inequality $\det(\text{Cov}_D) \geq \det(\text{qCov}_{D,f}^{as})$.

Which was based on numerous partial results for very specific f functions for few (generally 2) observables and the inequalities were expressed in different form. The conjecture was proved by Andai and Gibilisco, Imparato and Isola in 2008.

We have found a more accurate inequality:

Theorem [Lovas, Andai]. For any operator monotone function $f \in \mathcal{F}_{\text{op}}$ at every state $D \in \mathcal{M}_n^1$ for every N -tuple of observables $(A^{(k)})_{k=1, \dots, N}$ we have for the covariance matrices

$$\det(\text{Cov}_D) \geq \det(\text{qCov}_{D,f}^s) \geq \det(\text{qCov}_{D,f}^{as}).$$

We have an estimation for the gap between the symmetric and antisymmetric covariance:

Theorem [Lovas, Andai]. Using the same notation as in the previous Theorem we have

$$\det(\text{qCov}_{D,f}^s) - \det(\text{qCov}_{D,f}^{as}) \geq (2f(0))^N \det(\text{Cov}_D^{f_{\text{RLD}}}),$$

where $f_{\text{RLD}}(x) = \frac{2x}{1+x}$.

Moreover we have shown that the symmetric covariance generated by the function $f_{\text{opt}}(x) = \frac{1}{2} \left(\frac{1+x}{2} + \frac{2x}{1+x} \right)$ is universal in the following sense:

Theorem [Lovas, Andai]. For every function $g \in \mathcal{F}_{\text{op}}$ the inequality

$$\det(\text{qCov}_{D,f_{\text{opt}}}^s) \geq \det(\text{qCov}_{D,g}^{as})$$

holds and f_{opt} gives the best upper bound in \mathcal{F}_{op} .