Introduction: Decision theory in Classical Statistics, Sufficient Statistic and Rao-Quantum Theory of Sufficient Statistic and Decision Concluding Remarks

Sufficient Statistic, Rao-Blackwell Theorem in Quantum Probability

Kalyan B. Sinha

Indian Statistical Institute,

Indian Institute of Science,

&

J. N. Centre For Advanced Scientific Research, Bangalore, India

QP-42 Conference, Indian Statistical Institute, Bangalore, India

January 17-20, 2022

Overview

Introduction: Decision theory in Classical Statistics, Sufficient Statistic and Rao-Blackwell Theorem

2 Quantum Theory of Sufficient Statistic and Decision



Abraham Wald founded the Decision Theory in Classical Statistics:

Wald, Abraham.: "Statistical Decision Functions ", John Wiley & Sons, Inc., New York, N. Y.; Chapman & Hall, Ltd., London, 1950.

For a more recent account one can see the following:

 Ferguson, Thomas S.: "Mathematical Statistics-A Decision Theoretic Approach", Probability and Mathematical Statistics, Vol. 1, Academic Press, New York-London, 1967. The theory (or equivalently von Neumann's theory of games) starts with three basic objects:

(i) $\Theta,$ the parameter space for the (joint-) probability distribution or state of the random variables or observables.

(ii) Ω (the set of decisions), a measure space with $D : \mathcal{X} \times (\text{mble subset of } \Omega) \mapsto \mathbb{R}_+$, a family of probability measures, with \mathcal{X} being the sample(experiment) space.

(iii) $L: \Theta \times \Omega \to \mathbb{R}_+$, Loss function. Given $\{\Theta, (\Omega, D), L\}$, the Risk function

$$\mathcal{R}(\theta, D) \equiv \int_{\Omega} L(\theta, \omega) \int_{\mathcal{X}} \mu_{\theta}(dx) D(x, d\omega)$$

=
$$\int_{\Omega} L(\theta, \omega) (\mu_{\theta} \circ D) (d\omega)$$
(1)
=
$$\int_{\mathcal{X}} \mu_{\theta}(dx) \left(\int_{\Omega} L(\theta, \omega) D(x, d\omega) \right) \equiv \mathbb{E}_{\theta} \left(L(\theta, D(X)) \right),$$
(2)

where X stands for random variable in \mathcal{X} , and

$$\mathbb{E}_{ heta}(\cdot) = \int_{\mathcal{X}} (\cdot) \mu_{ heta}(dx).$$

The associated Baysian Risk function

$$\mathcal{R}(\pi,D)\equiv\int_{\Theta}\pi(d heta)\mathcal{R}(heta,D),$$

where π the 'prior' probability measure.

Goal of the Dicision theory:

Minimax Property, viz. Does there exist a "Decision rule D_0 " such that

 $\inf_{D \in \{D\}} \sup_{\Theta} \mathcal{R}(\theta, D) = \sup_{\Theta} \mathcal{R}(\theta, D_0) ?$

(3)

Among the set of random variables in \mathcal{X} , that is the set of real-valued measurable functions on \mathcal{X} , often one can find (in most cases due to existence of specific symmetries in the probability distributions involved) a small subset (often only one) called "sufficient statistic" \mathcal{T} , such that the "conditional probability given $\mathcal{T} = t$ " is independent of the parameter $\theta \in \Theta$. Equivalently, the joint probability distribution factorizes into

$$\phi(t,\theta) \cdot \rho_t \tag{4}$$

- $\rho_t = \text{conditional probability given equals to } T = t$,
- ϕ : (range of T) $\times \Theta \mapsto \mathbb{R}_+$ is the rest.

Example (Binomial):

Two i.i.d. $\{1,0\}$ - valued Random variables X_1, X_2 having binomial *p*-distribution:

$$Prob\{X_j = x_j\} = p^{x_j}(1-p)^{1-x_j}$$

For 0 , the joint p.d.f. is given by

$$Prob\{X_1 = x_1, X_2 = x_2\} = p^{(x_1 + x_2)}(1 - p)^{1 - (x_1 + x_2)}$$

Choose $T = X_1 + X_2$, then $Prob\{X_1 = x_1, X_2 = x_2\} = \phi(t, p)\rho_t$, where

$$\phi(t,p) = \begin{pmatrix} 2 \\ t \end{pmatrix} p^t (1-p)^t \text{ and }
ho_t = \begin{pmatrix} 2 \\ t \end{pmatrix}^{-1}$$

is the probability conditional to T = t.

Introduction: Decision theory in Classical Statistics, Sufficient Statistic and Rao

Quantum Theory of Sufficient Statistic and Decision Concluding Remarks

Here the symmetry is that of the transformation:



This gives two parallel definitions of sufficient statistic T, the first in terms of the conditional probability, given T = t, being independent of parameter θ or the joint probability distribution has the "factorization property (4)". It is the second which we adopt in Quantum case.

Theorem 1.1 (Classical Rao-Blackwell).

Let Ω be a convex subset of \mathbb{R}^n , $L(\theta, \cdot)$ be a convex function on Ω and let T be a sufficient statistic for μ_{θ} in (1). Then $\mathcal{R}(\theta, D) \geq \mathcal{R}(\theta, D_T)$, where $D_T(t) = \mathbb{E}(D(X)|T = t)$. Furthermore, $\mathbb{E}_{\theta}(D_T) = \mathbb{E}_{\theta}(D(X))$.

Proof.

Since $L(\theta, \cdot) : \Omega \to \mathbb{R}_+$ is convex, then by Jensen's inequality

$$\begin{split} \mathcal{R}(\theta,D) &= \mathbb{E}_{\theta} \left(L(\theta,D(X)) \right) = \mathbb{E}_{\theta} \left\{ \mathbb{E} \left(L(\theta,D(X)) | T = t \right) \right\} \\ &\geq \mathbb{E}_{\theta} \left\{ L(\theta,\mathbb{E}(D(X)|T = t) \right\} \\ &= \mathbb{E}_{\theta} \left(L(\theta,D_{T}(\cdot)) \right) = \mathcal{R}(\theta,D_{T}). \end{split}$$
$$\begin{split} \mathcal{E}_{\theta}(D_{T}) &= \mathbb{E}_{\theta} \left(\mathbb{E} \left(D(X) | T = t \right) \right) = \mathbb{E}_{\theta} \left(D(X) \right). \end{split}$$

D(X) is an unbiased(random) estimator for $\theta \implies$ the same for non-random D_T .

As mentioned earlier, we adapt the factorization property of a 'state' as in (4) to be the definition/ starting point for the Quantum theory.

Let

- $\underline{A} = (A_1, A_2, \dots, A_n)$ be a family of bounded commuting self-adjoint operators (simultaneously measurable observables) in \mathcal{H} .
- $T = T(\underline{A})$ be a self-adjoint operator function of \underline{A} with the unique spectral transformation (or diagonalizing unitary map):

$$\mathcal{U}:\mathcal{H}
ightarrow\int_{sp(T)}^{\oplus}\mathfrak{h}_t\
u(dt),$$

$$\circ \langle f,g
angle_{\mathcal{H}} = \int
u(dt) \langle f_t,g_t
angle_{\mathfrak{h}_t},$$
 and

• For
$$B \in {\underline{A}}'$$
,

 $(\mathcal{U}Bf)(t) = B(t)f_t, \ a.a.(\nu) \ t \in sp(T) \ \text{with} \ B(t) \in \mathcal{B}(\mathfrak{h}_t).$

Definition 2.1.

Let T be as above in \mathcal{H} and $\{\tau_{\theta}\}_{\theta\in\Theta}$ be a parameterized family of states on $\{\underline{A}\}'$. Then T is a "quantum sufficient statistic" (QSS) for $\{\tau_{\theta}\}$ if for every $B \in \{\underline{A}\}'$,

$$\tau_{\theta}(B) = \int \nu(dt)\phi(t,\theta) \operatorname{Tr}_{t}(B(t)\rho_{t}), \qquad (5)$$

where $\rho_t \in \mathcal{B}_{1+}(\mathfrak{h}_t)$, Tr_t is the $\mathcal{B}(\mathfrak{h}_t)$ -trace and $\phi : sp(T) \times \Theta \to \mathbb{R}_+$, a bounded measurable function, satisfying

$$\tau_{\theta}(I) = \int \nu(dt)\phi(t,\theta) \operatorname{Tr}_{t}(\rho_{t}) = 1$$
(6)

Remark 2.2.

 τ_{θ} is implemented by a density Matrix $\rho_{\theta} \stackrel{(5)}{\Longrightarrow} \rho_{\theta}$ commutes with T. Moreover, $\rho_{\theta} \in \mathcal{B}_1(\mathcal{H}) \implies$ the measure ν in (5) have a non-trivial atomic part.

In the Definition 2.1, the operator $\rho_t \in \mathcal{B}_{1+}(\mathfrak{h}_t)$ may be considered as the "conditional density matrix, given T = t," (except that $\operatorname{Tr}_t(\rho_t)$ non-negative, may not be equal to 1 for a.a. $(\nu)t$).

Note that, if $S = \{t \in sp(T) | \operatorname{Tr}_t(\rho_t) = 0\}$, then in the RHS of (5) and (6), the integrals has non-zero contributions only from $S^c \cap sp(T)$. Therefore, WLOG, we can assume that $S \cap sp(T) = \emptyset$. The earliest and simplest Quantum version of the decision theory (due to A.S. Holevo) the parameter space Θ , Decision space Ω remain classical (i.e. measure spaces), but

- the sampling/experiment space $\mathcal{X} \mapsto$ a separable Hilbert space \mathfrak{h}_s
- D: σ- algebra of measurable sets → positive operator-valued measure (POVM) in h_s satisfying
- Countable additivity:

$$D\left(igcup_1^\infty\Delta_j
ight) = \sum_{j=1}^\infty D(\Delta_j), ext{ converging in WOT}$$

 $D(\Omega) = I.$

 \implies Partially Quantum Risk function

$$\mathcal{R}(\theta, D) = \int_{\Omega} L(\theta, \omega) \ \tau_{\theta} \big(D(d\omega) \big), \tag{7}$$

 $\{\tau_{\theta} | \theta \in \Theta\}$ states on $\{\underline{A}\}'$, replaces the probability measure μ_{θ} in (1). The associated partially quantum Baysian risk function (with prior probability π):

$$\mathcal{R}(\pi, D) = \int_{\Theta} \pi(d\theta) \mathcal{R}(\theta, D)$$
$$= \int_{\Theta \times \Omega} L(\theta, \omega) \left\{ \pi(d\theta) \tau_{\theta} (D(d\omega)) \right\}$$
(8)

Combining Definition 2.1 and equations (7)-(8), leads to partially quantum Rao-Blackwell theorem.

- Holevo, A. S.: "Statistical decision theory for quantum systems", J. Multivariate Anal., **3** (1973), 337–394.
- Petz, D.: "Quantum information theory and quantum statistics", Theoretical and Mathematical Physics. Springer-Verlag, Berlin, 2008.
- Sinha, K. B.: "A Decision Theory in Non-Commutative Domain", Statistics and Applications, 19 (1), 1–8, 2021.

Theorem 2.3.

Let $\{\Theta, \Omega, L, D\}$ be the objects introduced earlier leading to expressions (7)-(8). Let Ω be a bounded convex subset of \mathbb{R}^n , let T be a QSS relative to $\{\tau_{\theta}\}$ and let $\{D(\cdot)\} \in \{\underline{A}\}'$. If furthermore, $L(\theta, \cdot)$ is convex, then

$$\mathcal{R}(\theta, D) \geq \mathcal{R}(\theta, D_T) \equiv \tau_{\theta}(L(\theta, D_T)),$$

where D_T is a unique bounded self-adjoint operator function of the QSS T. Moreover,

$$au_{ heta}(D_{T}) = au_{ heta}\left(\int_{\Omega} \omega D(d\omega)\right).$$

To prove the above theorem we need following two lemmas:

Lemma 2.4.

Let A be a bounded self-adjoint operator in \mathcal{H} and let σ be a density matrix. Then for $f : \mathbb{R} \to \mathbb{R}$ bounded measurable convex function,

 $\operatorname{Tr}(\sigma f(A)) \geq f(\operatorname{Tr}(\sigma A)).$

The proof of the above lemma follows easily from the spectral theorem and Jensen's inequality.

Lemma 2.5.

Let $\{D(\Delta)\}$ be a POVM in \mathcal{H} over Ω and let the family commute with a self-adjoint operator T. Then in the spectral representation of T given earlier, for a.a. $(\nu)t \in sp(T)$, the family $\{D_t(\Delta)|\Delta \text{ measuarble subset of }\Omega\}$ is a POVM in the decomposition

$$\mathfrak{h}_{s}=\int_{s\rho(T)}^{\oplus}\mathfrak{h}_{t}\nu(dt).$$

Sketch of proof of Lemma 2.5.

• Since $\langle f, D(\cdot)f \rangle$ is countably based, it suffices to look at the appropriate countable family $\{\Delta_k\}_{k=1}^{\infty}$ and

$$\langle f, D(\Delta_k)f \rangle = \int \nu(dt) \langle f_t, D_t(\Delta_k)f_t \rangle,$$

where $D_t(\Delta_k)$ is defined for $t \in N(\Delta_k)^c$ with $\nu(N(\Delta_k)) = 0$.

• Let $\{\Delta_j\}_1^\infty$ be a family of disjoint mble subsets of Ω and let δ be a ν - mble subset of sp(T). Then

$$\left\langle f, D\left(\bigcup_{j=1}^{\infty} \Delta_{j}\right) \mathcal{X}_{\delta}(T)g\right\rangle = \int_{\delta} \nu(dt) \sum_{j=1}^{\infty} \langle f_{t}, D_{t}(\Delta_{j})g_{t} \rangle_{t}$$
$$= \int_{\delta} \nu(dt) \left\langle f_{t}, D_{t}\left(\bigcup_{j=1}^{\infty} \Delta_{j}\right)g_{t} \right\rangle_{t}$$

 \implies Countable additivity for a.a.(ν) t.

Set
$$N = \bigcup_{k=1}^{\infty} N(\Delta_k)$$
, then $\nu(N) = 0$ and $D_t(\cdot)$ is defined $a.a.(\nu)t$.

Sketch of proof of Theorem 2.3

Definition 2.1 and equation (6) implies that

$$\mathcal{R}(\theta, D) = \int_{sp(T)} \nu(dt)\phi(t, \theta) \int_{\Omega} L(\theta, \omega) \operatorname{Tr}_{t} \left(\rho_{t} D_{t}(d\omega) \right)$$

- Observe that $\operatorname{Tr}_t(\rho_t) = 0 \implies \rho_t = 0$ since $\rho_t \ge 0$.
- \implies $S \equiv \{t \in sp(T) | \operatorname{Tr}_t(\rho_t) = 0\}$
- \implies the t- integral in (6) and the above integral is restricted to $S^c \cap sp(T)$, in which case $\rho_t > 0$

•
$$\implies$$
 For $t \in S^c \cap sp(T)$, set $\sigma_t = \left(\operatorname{Tr}_t(\rho_t)\right)^{-1} \rho_t$, then σ_t is a density matrix in \mathfrak{h}_t and by Lemma 2.4, since $L(\theta, \cdot)$ is convex

Continue...

 $\bullet \implies$

$$\int_{\Omega} L(\theta, \omega) \operatorname{Tr}_{t} \left(\sigma_{t} D_{t}(d\omega) \right) \geq L \left(\theta, \int_{\Omega} \omega \operatorname{Tr}_{t} \left(\sigma_{t} D_{t}(d\omega) \right) \right)$$
$$\Longrightarrow \mathcal{R}(\theta, D) \geq \int_{sp(T)} \nu(dt) \phi(t, \theta) \operatorname{Tr}_{t}(\rho_{t}) L(\theta, D_{T}(t)),$$

where

$$D_{\mathcal{T}}(t) = \begin{cases} \left(\mathsf{Tr}_t(\rho_t) \right)^{-1} \left(\int_{\Omega} \omega \, \mathsf{Tr}_t \left(\sigma_t D_t(d\omega) \right) \right), & t \in S^c \cap \mathfrak{sp}(\mathcal{T}) \\ 0, & t \in S. \end{cases}$$

• $D_{T}(\cdot)$ is a bounded measurable function, defines a bounded operator, commuting with QSS T, and

$$\mathcal{R}(\theta, D) \geq \tau_{\theta} \Big(L(\theta, D_T) \Big) = \mathcal{R}(\theta, D_T).$$

Continue...

• Furthermore,

$$\begin{split} \theta &\equiv \tau_{\theta} \left(\int_{\Omega} \omega D(d\omega) \right) = \int_{sp(T)} \nu(dt) \phi(t,\theta) \operatorname{Tr}_{t} \left(\rho_{t} \left(\int_{\Omega} \omega D_{t}(d\omega) \right) \right) \\ &= \int_{sp(T)} \nu(dt) \phi(t,\theta) \operatorname{Tr}_{t} \left(\rho_{t} \right) D_{T}(t) \\ &= \tau_{\theta} (D_{T}), \end{split}$$

that is the expectation of "Rao-Blackwell observable" D_T in state τ_{θ} is an unbiased estimator for the parameter θ .

Corollary 2.6 (Partially quantum Baysian Rao-Blackwell theorem).

Under the set of hypothesis of Theorem 2.3, the risk function

$$\mathcal{R}(\pi,D) \geq \mathcal{R}(\pi,D_{\mathcal{T}}) = \int_{\Theta} \pi(d heta) \mathcal{R}(heta,D_{\mathcal{T}}).$$

Furthermore,

$$\int \pi(d\theta)\theta = \int \pi(d\theta)\tau_\theta\left(\int \omega D(d\omega)\right) = \int \pi(d\theta)\tau_\theta(D_T).$$

Proof of the above corollary immediate from Theorem 2.3.

Example 2.7.

Let
$$\mathcal{H} = L^2(\mathbb{R}^3)$$
, $\underline{P} = (P_1, P_2, P_3)$ -the momentum operators,
 $H_0 = \sum_{j=1}^3 p_j^2 = -\Delta$ (as self-adjoint operator in \mathcal{H}), $\underline{L} \equiv (L_1, L_2, L_3)$ -the
angular momentum operators with $\underline{L}^2 = \sum_{j=1}^3 L_j^2$, the total angular
momentum operators.

For the operator H_0 , the spectral representation is given by the unitary isomorphism via the Fourier transform : $\mathcal{H} \simeq L^2(\mathbb{R}_+; L^2(S^{(2)}), \frac{1}{2}t^{1/2}dt)$, with

$$(H_0 f)_t = t f_t \in \mathfrak{h}_t \simeq L^2(S^{(2)}) \ \forall \ t, \tag{9}$$

where $S^{(2)}$ is the unit sphere of 2-dimensions embedded in \mathbb{R}^3 .

On the other hand, we set $\tilde{\rho}_t = (\underline{L}^2 + 1)^{-2}$ in $\mathfrak{h}_t \forall t$ and note that $\tilde{\rho}_t$ is a positive trace-class operator in $\mathfrak{h}_t \cong L^2(S^{(2)})$ and its trace:

$$Tr_t(\widetilde{
ho}_t) = \sum_{\ell=0}^{\infty} rac{(2\ell+1)}{(\ell(\ell+1)+1)^2} \equiv C < \infty.$$

If we set $\varphi(t,\theta) = (\pi/16\theta^3)^{-1/2} \exp(-\theta t)$ and normalise $\rho_t = C^{-1} \tilde{\rho}_t$, then it is an easy calculation to verify that

$$\int \varphi(t,\theta)(Tr_t\rho_t)\frac{1}{2}t^{1/2}dt = 1.$$
(10)

Therefore we define the state τ_{θ} on $\mathcal{A} = \{H'_0\}$, the von Neumann algebra driven by the constants of free motion, generated by the Hamiltonian H_0 , as

$$\tau_{\theta}(B) = \int_{0}^{\infty} \varphi(t,\theta) \operatorname{Tr}_{t}(\rho_{t}B(t)) \cdot \frac{1}{2} t^{1/2} dt, \qquad (11)$$

setting the stage to recognize the self-adjoint operator $H_0 = T$, the sufficient statistic for this state τ_{θ} .

In this example, the symmetry group of rotations in 3-dimensions has played a role in the background.

Example 2.8.

The 2-dimensional (spin 1/2) representation of the proper rotation group $O^+(3)$ in \mathbb{R}^3 leads to the following standard operators in \mathbb{C}^2 : $\underline{S} = \{S_1, S_2, S_3\}, \underline{S}^2 = \sum_{j=1}^3 S_j^2$ and the spanning eigenbases for (\underline{S}^2, S_3) given by $|3/4, \uparrow\rangle$ and $|3/4, \downarrow\rangle$, in which \underline{S}^2 has 1/2(1/2 + 1) = 3/4 and S_3 has $\pm 1/2$ as eigenvalues respectively with \uparrow representing +1/2 and \downarrow for -1/2. Introduction: Decision theory in Classical Statistics, Sufficient Statistic and Rao-Quantum Theory of Sufficient Statistic and Decision Concluding Remarks

- Next consider in H = C² ⊗ C² ≃ C⁴, two such independent spin systems <u>S</u>^(j) (j = 1, 2) be given and we form the new observables: <u>S</u> = <u>S</u>⁽¹⁾ ⊗ I₂ + I₁ ⊗ <u>S</u>⁽²⁾ and corresponding <u>S</u>² in H.
- Then we look for the decomposition of *H*, corresponding to the irreducible representations of O⁺(3): *H* ≃ 𝔥₁ ⊕ 𝔥₀, where 𝔥₁ and 𝔥₀ are spanned by the eigenbases of {<u>S</u>², S₃} of the total system:

for
$$\mathfrak{h}_1: |2,\uparrow\uparrow\rangle, \ \left|2,\frac{\uparrow\downarrow+\downarrow\uparrow}{\sqrt{2}}\right\rangle, \ |2,\downarrow\downarrow\rangle \text{ and } \text{ for } \mathfrak{h}_0: \left|0,\frac{\uparrow\downarrow-\downarrow\uparrow}{\sqrt{2}}\right\rangle,$$
(12)

in which the normalized eigenstates corresponding to the eigenvalues (1, 0, -1) in \mathfrak{h}_1 and (0) in \mathfrak{h}_0 of S_3 respectively are represented pictorially as above.

- Let the state be given by $\rho_{\theta} = Z^{-1} \exp(-\theta \underline{S}^{(1)} \cdot \underline{S}^{(2)})$, where Z is such that $Tr \rho_{\theta} = 1$.
- An easy calculation, using the identity: $\underline{S}^{(1)} \cdot \underline{S}^{(2)} = 1/2(\underline{S}^2 - \underline{S}^{(1)^2} - \underline{S}^{(2)^2})$ (leads to the fact that the operator $(\underline{S}^{(1)} \cdot \underline{S}^{(2)})$ has eigenvalues 1/4 and -3/4 respectively in each eigenfamilies in \mathfrak{h}_1 and \mathfrak{h}_0 respectively.
- Thus $\rho_{\theta} = Z^{-1}(e^{-\theta/4}P_1 + e^{3\theta/4}P_0)$, where P_1 and P_0 are the projections onto $\mathfrak{h}_1(3 \dim)$ and $\mathfrak{h}_0(1 \dim)$ respectively, in \mathcal{H} .
- This implies that $Z = Tr(e^{-\theta/4}P_1 + e^{3\theta/4}P_0) = 3e^{-\theta/4} + e^{3\theta/4}$, which leads to the following expression for ρ_{θ} :

$$\rho_{\theta} = (3 + e^{\theta})^{-1} P_1 + (1 + 3e^{-\theta})^{-1} P_0.$$
(13)

Introduction: Decision theory in Classical Statistics, Sufficient Statistic and Rao-Quantum Theory of Sufficient Statistic and Decision Concluding Remarks

• Therefore, here $\underline{S}^2 = (\underline{S}^{(1)} + \underline{S}^{(2)})^2$ is the candidate for "quantum sufficient statistic" and

$$\rho_{\theta}(B) = \sum_{s=0}^{1} \varphi(s,\theta) \operatorname{Tr}_{s}(\rho_{s}B(s)),$$

where s(s+1) are the eigenvalues of \underline{S}^2 , and

$$\varphi(0,\theta) = (1+3e^{-\theta})^{-1}, \varphi(1,\theta) = (3+e^{\theta})^{-1};$$

 $\rho_0 = P_0$, $\rho_1 = P_1$, the orthogonal projections respectively.

Fully Quantum Rao-Blackwell theorem:

For this part, we shall assume that the QSS T is a bounded self-adjoint operator in \mathfrak{h}_s (the Hilbert space of observations) with only discrete spectrum.

Unlike in the previous section, here the Baysian part is quantized \implies the Baysian Hilbert space \mathfrak{h}_B and the theory is put in $\mathcal{H} = \mathfrak{h}_B \times \mathfrak{h}_s$, Φ is a density Matrix in \mathcal{H} , $\Omega = [a, b] \subseteq \mathbb{R}$, $L : \Omega \to \mathcal{B}_+(\mathfrak{h}_B)$ strongly continuous, $\{D(\cdot)\}$ POVM commuting with QSS T in \mathfrak{h}_s .

Then one has

 $L \cdot D \equiv \int_a^b L(\omega) D(d\omega)$ exists as trong Riemann-Stiltje's integral on \mathcal{H} ,

 $\|L \cdot D\|_{\mathcal{B}(\mathcal{H})} \leq \sup_{\omega} \|L(\omega)\|_{\mathcal{B}(\mathfrak{h}_B)}.$

The fully quantum risk function:

$$\mathcal{R}(\Phi, D) = \Phi(L \cdot D)$$

Here

$$\mathcal{H}\simeq \mathfrak{h}_B\otimes \int^\oplus
u(dt)\mathfrak{h}_t\simeq \int^\oplus
u(dt)(\mathfrak{h}_B\otimes \mathfrak{h}_t),$$

 ν is only atomic and $\Phi \simeq \int^{\oplus} \nu(dt)(\Phi_B(t) \otimes \rho_t), \ \Phi_B(t) \in \mathcal{B}_{1+}(\mathfrak{h}_B),$

$$\operatorname{Tr} \Phi = 1 \int \nu(dt) \operatorname{Tr}_B(\Phi_B(t)) \operatorname{Tr}(\rho_t).$$

Fully quantum Rao-Blackwell Theorem II

Assume all that has gone before, and $S = \{t \in sp(T) | \rho_t = 0\}$ is a mble set and set $D_T(t)$ as before to get a mble bounded function. Let L be weakly convex, that is. $\omega \mapsto \langle f, L(\omega)f \rangle$ is a positive convex, continuous function for every $f \in \mathfrak{h}_B$. Then

 $\Phi(L \cdot D) \geq \Phi(L(D_T)),$

where $L(D_T)(t) = L(D_T(t))$ and range of $D_T(\cdot) \subseteq \Omega$.

$$\Phi\left(\int_{\Omega} \omega D(d\omega)\right) = \int \nu(dt) \operatorname{Tr}_{B}(\Phi_{B}(t)).$$

$$\operatorname{Tr}\left(\rho_{t} \int \omega D_{t}(d\omega)\right) = \int \nu(dt) \operatorname{Tr}\left(\Phi_{B}(t) \otimes \rho_{t}\right) D_{T}(t) = \Phi(D_{T}).$$

More general theory needs the use of the central decomposition of von Neumann algebras and of states on them.

