

HOW SHARP ARE PV MEASURES?

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Properties of sharp observables (normalized PV measures) in relation to smearing by a Markov kernel are studied. Conditions under which a sharp observable is a smearing of another, in general unsharp observable, are found.

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1. Introduction

Normalized POV (positive operator valued) measures are used to describe generalized observables in quantum mechanics [17, 13, 5]. Their introduction is justified by the analysis of some ideal experiments which shows that there are quantum events that cannot be described by projections [5]. POV measures are also used to generalize Mackey's imprimitivity theorem [2, 21] and to study the problem of the joint measurements of incompatible observables [18, 22, 6, 24].

Generalized yes-no experiments are in one-to-one correspondence with self-adjoint operators lying between 0 and I (with respect to the usual ordering of self-adjoint operators). These operators are called *quantum effects*. Let $\mathcal{E}(H)$ denote the set of all quantum effects on a Hilbert space H , i.e. $\mathcal{E}(H) := \{a : 0 \leq a \leq I\}$, where a is a self-adjoint operator. Projection operators are contained in $\mathcal{E}(H)$, and they are distinguished among the effects by the equality $p \wedge (I - p) = 0$, which can be interpreted as the property that events p and non- p cannot simultaneously occur. Projection operators are called *sharp* effects, while the other effects are *unsharp*. Correspondingly, PV (projection valued) observables are called *sharp* observables [1].

Recall that the states on $\mathcal{E}(H)$ (i.e. the physical states of the corresponding physical system) coincide with the set of all density operators on H . There exists a one-to-one correspondence between POV measures (defined on a measurable space (X, \mathcal{B})) and affine maps from the set of states into the set of probability measures

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on (X, \mathcal{B}) , which is based on the interpretation of the number $\text{Tr}[mF(A)]$ as the probability that the outcome of a measurement of the observable (POV measure) F is in $A \in \mathcal{B}$ if the physical system is in the state m [13]. This one-to-one correspondence allows one to apply some results of the classical mathematical statistics to quantum experiments [13, 16]. In particular, given a probability measure μ and a suitable Markov kernel λ , we can form another probability measure, $\lambda \circ \mu$, so-called randomization of μ by λ [20]. This has been applied to quantum observables: to a given observable and a suitable Markov kernel, a new observable can be created, which is called a smearing, or a fuzzy version, of the given observable [11–13, 16]. For example, it is well-known that an unsharp observable is a smearing of a sharp observable iff its range is commutative [14, 3, 7, 16]. A partial ordering can be introduced on the set of observables by defining $E \leq F$ if the observable F is a smearing of the observable E [4, 10, 16]. Minimal points in this ordering are called clean observables [4].

In the present paper, we study properties of sharp observables in relation to smearing. In our considerations, we often replace a Markov kernel by a weak Markov kernel to simplify the proofs, and then apply well known results about the equivalence of the weak Markov kernel with its regular version, which is a Markov kernel. We show that a sharp observable P , defined on a standard Borel space, can be considered as a smearing of another, in general unsharp observable M , iff the corresponding Markov kernel is of a special type, which makes the sharp observable P a function of the unsharp observable M . We also show that this holds not only for sharp observables, but for all observables which are extremal with respect to the convex structure of observables. Consequently, a sharp observable is clean iff its range generates a maximal abelian von Neumann subalgebra of the bounded operators on H . We also show that for a sharp observable P defined on a standard Borel space, and an arbitrary observable M , the following properties are equivalent: (a) the range of P is contained in the range of M ; (b) P is a function of M ; (c) P is a smearing of M . We note that the equivalence of (a) and (b) has been proved in [8], where the Naimark theorem was used. In this paper, we give a different proof.

2. Smearing of observables

Let H be a (complex, separable) Hilbert space. Let $\mathcal{E}(H)$ be the set of effects on H and let \mathcal{S} be the set of states on $\mathcal{E}(H)$. We recall the following property of the order on $\mathcal{E}(H)$, inherited from the usual order on self-adjoint operators:

$$a \leq b \quad \text{if and only if} \quad ab = ba = a \quad (1)$$

whenever $a, b \in \mathcal{E}(H)$ and a or b is a projection.

Let (X, \mathcal{A}) and (Y, \mathcal{B}) be measurable spaces and let $E : (X, \mathcal{A}) \rightarrow \mathcal{E}(H)$ be a POV measure. Assume further that there is a map $\lambda : X \times \mathcal{B} \rightarrow [0, 1]$ such that

- (i) $\lambda(\cdot, B)$ is \mathcal{A} -measurable for all $B \in \mathcal{B}$,
- (ii) $\lambda(x, \cdot)$ is a probability measure on \mathcal{B} for all $x \in X$.

That is, λ is a Markov kernel. Then

$$\lambda \circ E(B) := \int_X \lambda(x, B) E(dx), \quad B \in \mathcal{B}, \quad (2)$$

defines a POV measure $(Y, \mathcal{B}) \rightarrow \mathcal{E}(H)$, called the *smearing* of E with respect to λ .

The notion of a Markov kernel can be weakened as follows. Let $\mathcal{P} \subseteq M_1^+(X, \mathcal{A})$, where $M_1^+(X, \mathcal{A})$ denotes the set of probability measures on (X, \mathcal{A}) , and let $\nu : X \times \mathcal{B} \rightarrow \mathbb{R}$. We will say that ν is a *weak Markov kernel with respect to \mathcal{P}* if:

- (i) $x \mapsto \nu(x, B)$ is \mathcal{A} -measurable for all $B \in \mathcal{B}$;
- (ii) for every $B \in \mathcal{B}$, $0 \leq \nu(x, B) \leq 1$, \mathcal{P} -a.e.;
- (iii) $\nu(x, Y) = 1$, \mathcal{P} -a.e. and $\nu(x, \emptyset) = 0$, \mathcal{P} -a.e.
- (iv) if $\{B_n\}$ is a sequence in \mathcal{B} such that $B_n \cap B_m = \emptyset$ for $m \neq n$, then

$$\nu(x, \bigcup_n B_n) = \sum_n \nu(x, B_n), \quad \mathcal{P}\text{-a.e.}$$

Let E be as above and put $\mathcal{P} = \{m \circ E : m \in \mathcal{S}\}$. If $\nu : X \times \mathcal{B} \rightarrow \mathbb{R}$ is a weak Markov kernel with respect to \mathcal{P} , then we will say that ν is a weak Markov kernel with respect to E and

$$\nu \circ E(B) := \int_X \nu(x, B) E(dx), \quad B \in \mathcal{B},$$

defines a POV measure, which will be called a *smearing* of E with respect to ν .

REMARK 2.1. We note that a weak Markov kernel $\nu : X \times \mathcal{B} \rightarrow [0, 1]$ (with respect to one probability measure P) is called a *random measure* in the literature. If \mathcal{B} is the Borel σ -algebra of subsets of a complete separable metric space Y , then there exists a regular version ν^* of ν , such that ν^* is a Markov kernel, and

$$\forall B \in \mathcal{B}, \nu(x, B) = \nu^*(x, B), \quad P\text{-a.e.} \quad (3)$$

(see, e.g. [20, VI.1. 21.]).

Notice further that since on a separable Hilbert space there exists a faithful state $m_0 \in \mathcal{S}$, and $m \circ E$ is dominated by $m_0 \circ E$ for all $m \in \mathcal{S}$, then ν is a weak Markov kernel with respect to $\{m \circ E : m \in \mathcal{S}\}$ iff ν is a weak Markov kernel with respect to $m_0 \circ E$.

Moreover, it has been proved in [16] that if an observable $F : (Y, \mathcal{B}) \rightarrow \mathcal{E}(H)$ is a smearing of an observable E with respect to a weak Markov kernel ν , and (Y, \mathcal{B}) is a standard Borel space, then there is a Markov kernel ν^* such that F is a smearing of E with respect to ν^* .

3. PV measures and smearings

For an observable E , let $\mathcal{R}(E)$ denote the range of E . The following theorem is well known, see [14, 3, 7, 16]. For completeness, we include (a sketch of) the proof, as it was given in [16].

THEOREM 3.1. *Let $M : (Y, \mathcal{B}) \rightarrow \mathcal{E}(H)$ be a POV measure. Then M is a smearing of some PV measure P with respect to a weak Markov kernel if and only if $\mathcal{R}(M)$ is commutative.*

Proof: Since $\mathcal{R}(M)$ is commutative, there is a self-adjoint operator T on H , such that $\mathcal{R}(M) \subset \{T\}''$. It follows that for each $B \in \mathcal{B}$, there is a Borel function f_B , such that

$$M(B) = f_B(T) = \int_{\mathbb{R}} f_B(x) P(dx),$$

where P is the spectral measure of T . It is not difficult to show that $\nu(x, B) = f_B(x)$ defines a weak Markov kernel $X \times \mathcal{B} \rightarrow \mathbb{R}$ with respect to P .

The converse statement is obvious. \square

The main purpose of this paper is to study the opposite situation, namely when a PV measure P is a smearing of some observable M .

THEOREM 3.2. *Let $M : (X, \mathcal{A}) \rightarrow \mathcal{E}(H)$ be a POV measure and let $P : (Y, \mathcal{B}) \rightarrow \mathcal{E}(H)$ be a PV measure. Let $\nu : X \times \mathcal{B} \rightarrow \mathbb{R}$ be a weak Markov kernel with respect to M and suppose that $P = \nu \circ M$. Then $\mathcal{R}(P) \subset \mathcal{R}(M)$.*

Proof: Let $B \in \mathcal{B}$, then $P(B) = \int \nu(x, B) M(dx)$. Put $\pi(B) := \{x : \nu(x, B) = 1\}$. Let m be a state on $\mathcal{E}(H)$, with the support $\text{supp}(m) = P(B)$. Then

$$1 = m(P(B)) = \int \nu(x, B) m(M(dx)),$$

hence $m \circ M(\pi(B)^c) = 0$. Since $P(B)$ is the support of m , we have

$$P(B)M(\pi(B)^c)P(B) = 0.$$

By positivity of M this entails

$$P(B)M(\pi(B)^c) = M(\pi(B)^c)P(B) = 0.$$

Therefore $P(B)M(\pi(B)) = M(\pi(B))P(B) = P(B)$, hence $P(B) \leq M(\pi(B))$. Similarly, $P(B^c) \leq M(\pi(B^c))$. But

$$I = P(B) + P(B^c) = M(\pi(B)) + M(\pi(B)^c)$$

yields $M(\pi(B)^c) \leq P(B^c)$, and since, by definition, $\pi(B^c) \subseteq \pi(B)^c$ modulo M , we get

$$P(B^c) \leq M(\pi(B^c)) \leq M(\pi(B)^c) \leq P(B^c).$$

We conclude that $P(B^c) = M(\pi(B^c)) = M(\pi(B)^c)$, and therefore $P(B) = M(\pi(B)) \in \mathcal{R}(M)$. \square

As an example, we will consider in detail the case of a finite-dimensional Hilbert space.

EXAMPLE 3.1. Let H be finite dimensional. Let Y be a finite set and let $P : Y \rightarrow \mathcal{E}(H)$ be a PV measure. Assume that $P = \nu \circ M$ with a weak Markov

kernel ν and a POV measure $M : (X, \mathcal{A}) \rightarrow \mathcal{E}(H)$. Since Y is finite, there is a set $C \subset Y$, such that the restriction of ν to C^c is a Markov kernel and $M(C) = 0$.

For $y \in Y$, put $\pi(y) := \{x \in C^c : \nu(x, y) = 1\}$. As in Theorem 3.2, $P(y) = M(\pi(y))$. Moreover, since $\sum_{y \in Y} \nu(x, y) = 1$ for $x \in C^c$, we obtain that $x \in \pi(y)$ implies that $\nu(x, y') = 0$ and therefore $x \in \pi(y')^c$, for $y' \neq y$. This shows that $\{\pi(y) : y \in Y, C\}$ is a partition of Y .

Moreover, we can define a Markov kernel $\nu^* : X \times \mathcal{B} \rightarrow [0, 1]$ by

$$\nu^*(x, y) = \begin{cases} \chi_{\pi(y)}(x) & : x \in C^c, \\ \mu(y) & : x \in C, \end{cases}$$

where μ is any probability measure on Y . Then $\nu(x, y) = \nu^*(x, y)$ for $x \in C^c$ and we have $P = \nu^* \circ M$. The observable M has the following form: if $H_y := P(y)H$, then $H = \oplus_y H_y$ and

$$M(A) = \oplus_y P(y)M(A) = \oplus_y M(A \cap \pi(y)).$$

This can be interpreted as follows: measurement of the observable M consists of measuring the observable P and if the result is y , we measure the observable $P(y)MP(y)$, $y \in Y$. Measurements of this kind are considered, e.g. in [9].

In the above example, note that the weak Markov kernel must satisfy $\nu(x, y) \in \{0, 1\}$ for $y \in Y$ and all x in C^c . More generally, if $\nu : X \times \mathcal{B} \rightarrow \mathbb{R}$ is a weak Markov kernel with respect to a POV measure M , we will say that ν has values in $\{0, 1\}$ if for each $B \in \mathcal{B}$, $\nu(x, B) \in \{0, 1\}$, a.e. - $\{m \circ M, : m \in \mathcal{S}\}$.

Let (X, \mathcal{A}) be a measurable space, and let $E_i : (X, \mathcal{A}) \rightarrow \mathcal{E}(H)$, $i = 1, 2$, be POV measures. For every $\alpha \in [0, 1]$, $A \mapsto E(A) = \alpha E_1(A) + (1 - \alpha)E_2(A)$, $A \in \mathcal{A}$, defines a POV measure. Hence the set of all observables associated with (X, \mathcal{A}) bears a convex structure. Since projections are extremal points in the convex set $\mathcal{E}(H)$, sharp observables are extremal in the set of all observables associated with a given measurable space. In general, however, there exist extremal points which are unsharp [15].

THEOREM 3.3. *Let E be a POV measure which is an extreme point in the convex set of all POV measures defined on a measurable space (X, \mathcal{A}) . Then if E is a smearing of a POV measure M with respect to a weak Markov kernel ν , then ν has values in $\{0, 1\}$. Moreover, if (X, \mathcal{A}) is a standard Borel space, then E is a function of M .*

Proof: Let $M : (Y, \mathcal{B}) \rightarrow \mathcal{E}(H)$ be a POV measure and let $\nu : Y \times \mathcal{A} \rightarrow \mathbb{R}$ be a weak Markov kernel with respect to M . Suppose that $E = \nu \circ M$.

Fix $B_1 \in \mathcal{A}$. Define

$$\nu^\pm(y, B) := \nu(y, B) \pm [\nu(y, B_1)\nu(y, B \cap B_1^c) - \nu(y, B_1^c)\nu(y, B \cap B_1)].$$

Then $\nu^\pm(y, B)$ is a weak Markov kernel with respect to M [13]. Moreover,

$$\nu(y, B) = \frac{1}{2}\nu^+(y, B) + \frac{1}{2}\nu^-(y, B).$$

This implies that $E(B) = \frac{1}{2}E^+(B) + \frac{1}{2}E^-(B)$, where $E^\pm(B) = \int_Y v^\pm(y, B)M(dy)$. Since E is extremal, we must have $E^+ = E^- = E$, which implies that

$$\int [\nu(y, B_1)\nu(y, B \cap B_1^c) - \nu(y, B_1^c)\nu(y, B \cap B_1)]M(dy) = 0.$$

Then

$$\begin{aligned} E(B) &= \int \nu(y, B)M(dy) \\ &= \int (\nu(y, B \cap B_1) + \nu(y, B \cap B_1^c))M(dy) \\ &= \int [\nu(y, B_1)\nu(y, B \cap B_1) + \nu(y, B_1^c)\nu(y, B \cap B_1) + \nu(y, B \cap B_1^c)]M(dy) \\ &= \int (\nu(y, B_1)\nu(y, B \cap B_1) + \nu(y, B_1)\nu(y, B \cap B_1^c) + \nu(y, B \cap B_1^c))M(dy) \\ &= \int [\nu(y, B_1)\nu(y, B) + \nu(y, B \cap B_1^c)]M(dy). \end{aligned}$$

In particular,

$$E(B_1) = \int \nu(y, B_1)^2 M(dy) = \int \nu(y, B_1) M(dy),$$

and since $\nu(y, B_1) \geq \nu(y, B_1)^2$ a.e. M , we get $\nu(y, B_1) = \nu(y, B_1)^2$ a.e. M , that is, $\nu(y, B_1) \in \{0, 1\}$ a.e. M . Since B_1 was arbitrary, this holds for all $B \in \mathcal{A}$.

Suppose next that (X, \mathcal{A}) is a standard Borel space. Then there exists a Markov kernel $\lambda : Y \times \mathcal{A} \rightarrow [0, 1]$, such that $E = \lambda \circ M$. Put $\pi(A) := \{y : \lambda(y, A) = 1\}$.

First, we will show that $E(A) = M(\pi(A))$. For every A , $\pi(A) \cap \pi(A^c) = \emptyset$, because $\lambda(y, \cdot)$ is a probability measure. Therefore, there is a partition $Y = \pi(A) \cup \pi(A^c) \cup C_A$ and, by the first part of the proof, $M(C_A) = 0$. Then

$$E(A) = \int_{\pi(A)} \lambda(y, A)M(dy) + \int_{\pi(A^c)} \lambda(y, A)M(dy) + \int_{C_A} \lambda(y, A)M(dy).$$

The first integral is $M(\pi(A))$, the other two are 0.

Next, we show that π is a σ -homomorphism of sets modulo M .

- (1) $\pi(A \cap B) = \{y : \lambda(y, A \cap B) = 1\}$. Since $\lambda(y, \cdot)$ is a probability measure, we have $\lambda(y, A) = \lambda(y, B) = 1$ if and only if $\lambda(y, A \cap B) = 1$. This entails $\pi(A \cap B) = \pi(A) \cap \pi(B)$.
- (2) Observe that $A \subset B$ implies $\pi(A) \subset \pi(B)$, which follows from $\lambda(y, A) \leq \lambda(y, B)$.
- (3) Let $A_n \in \mathcal{A}$, $n \in \mathbb{N}$, $A_n \cap A_m = \emptyset$ if $n \neq m$. Put $A = \cup A_n$. Then $A_n \subset A \Rightarrow \pi(A_n) \subset \pi(A)$, hence $\cup \pi(A_n) \subset \pi(A)$. Let $C = \pi(A) \setminus \cup \pi(A_n)$. If $y \in C$, then $\lambda(y, A) = 1$ and $\lambda(y, A_n) \neq 1$ for all n , so that, for all n , either $\lambda(y, A_n) = 0$ or $y \in C_{A_n}$. Since $\lambda(y, A) = \sum \lambda(y, A_n) = 1$, there is

an n such that $\lambda(y, A_n) \neq 0$, that is, $y \in C_{A_n}$. Therefore $C \subseteq \bigcup_n C_{A_n}$, so that $M(C) = 0$. This concludes the proof that π is a set homomorphism modulo M .

Let m be a faithful state on $\mathcal{E}(H)$, $m \circ M = \mu$ is a probability measure on \mathcal{B} . Put $I := \{B \in \mathcal{B} : \mu(B) = 0\}$, then I is a σ -ideal, and \mathcal{B}/I is a Boolean σ -algebra. Let $p : \mathcal{B} \mapsto [\mathcal{B}]$ be the canonical homomorphism. Put $\pi_1 : \mathcal{A} \xrightarrow{\pi} \mathcal{B} \xrightarrow{p} \mathcal{B}/I$. Then π_1 is a σ -homomorphism. The triple (Y, \mathcal{B}, p) , where $p : \mathcal{B} \rightarrow \mathcal{B}/I$ is surjective, satisfies conditions of [19, Lemma 4.1.8], resp. [23, Theorem 1.4], and hence there is $f : Y \rightarrow X$, measurable and such that $\pi_1(A) = p \circ f^{-1}(A)$, $A \in \mathcal{A}$. By the definition of I , and since m is faithful, if $B_1 \in [B]$, then $M(B_1) = M(B)$. Hence $\pi_1(A) = [f^{-1}(A)] = [\pi(A)] \Rightarrow E(A) = M(\pi(A)) = M(f^{-1}(A))$. \square

Next we will show the converse to Theorem 3.2.

THEOREM 3.4. *Let $P : (Y, \mathcal{B}) \rightarrow \mathcal{E}(H)$ be a PV measure and let $M : (X, \mathcal{A}) \rightarrow \mathcal{E}(H)$ be a POV measure. If $\mathcal{R}(P) \subseteq \mathcal{R}(M)$, then there is a weak Markov kernel ν with respect to M , such that P is a smearing of M .*

Proof: The assumption implies that there is a mapping $\pi : \mathcal{B} \rightarrow \mathcal{A}$ such that $P(B) = M(\pi(B))$, $B \in \mathcal{B}$. The latter equality entails that

$$P(B) = \int_X \chi_{\pi(B)} M(dx), B \in \mathcal{B}.$$

Put $\nu(x, B) = \chi_{\pi(B)}(x)$, $B \in \mathcal{B}$. We will prove that $\nu : X \times \mathcal{B} \rightarrow [0, 1]$ is a weak Markov kernel with respect to M .

Clearly, $0 \leq \nu(x, B) \leq 1$, $\nu(x, Y) = 1$ a.e. M , $\nu(x, \emptyset) = 0$ a.e. M .

Let $\{B_n\}_n$ be a sequence of elements in \mathcal{B} , $B_m \cap B_n = \emptyset$, $m \neq n$, and denote $B := \bigcup_n B_n$. We have

$$M(\pi(B)) = P(B) = \sum_n P(B_n) = \sum_n M(\pi(B_n)).$$

We will show that

$$\sum_n M(\pi(B_n)) = M\left(\bigcup_n \pi(B_n)\right).$$

First, observe that $B_1 \cap B_2 = \emptyset$ implies that $M(\pi(B_1))M(\pi(B_2)) = 0$ and from $M(\pi(B_1) \cap \pi(B_2)) \leq M(\pi(B_1))M(\pi(B_2))$ we derive that

$$M(\pi(B_1) \cap \pi(B_2)) = 0.$$

Consider the sequence $\{C_n\}$, where

$$C_1 = \pi(B_1),$$

...

$$C_n = \pi(B_n) \setminus \bigcup_{k=1}^{n-1} \pi(B_k) = \bigcap_{k=1}^{n-1} \pi(B_n) \cap \pi(B_k)^c$$

...

Then $C_n \cap C_m = \emptyset$, $n \neq m$, and $C_n \subseteq \pi(B_n) \forall n$. In addition,

$$\begin{aligned} A_n &:= \pi(B_n) \cap C_n^c = \pi(B_n) \cap \left(\bigcup_{k=1}^{n-1} \pi(B_k)^c \cup \pi(B_k) \right) \\ &= \bigcup_{k=1}^{n-1} \pi(B_n) \cap \pi(B_k), \end{aligned}$$

which implies $\pi(B_n) = C_n \cup A_n$, $M(A_n) = 0$, whence $M(\bigcup_n \pi(B_n)) = M(\bigcup_n C_n) = \sum_n M(C_n) = \sum_n M(\pi(B_n)) = M(\pi(B))$.

We will show that we can replace C_n by a sequence D_n such that $D_n \subset \pi(B) \forall n$. Clearly, $P(B_n) \leq P(B)$ for all n . Since $M(\pi(B_n) \cap \pi(B)^c) \leq M(\pi(B_n)) = P(B_n) \leq P(B)$ and also $M(\pi(B_n) \cap \pi(B)^c) \leq M(\pi(B))^c = P(B)^c$, we obtain

$$M(\pi(B_n) \cap \pi(B)^c) = 0.$$

Since $C_n \subseteq \pi(B_n)$ we have $M(C_n \cap \pi(B)^c) = 0$. Put

$$D_n := C_n \cap \pi(B), \quad F_n := \pi(B_n) \setminus D_n = C_n \cap \pi(B)^c \cup A_n.$$

Then $\pi(B_n) = D_n \cup F_n$, with $M(F_n) = 0$. Moreover, $D_n \cap D_m = \emptyset$, $n \neq m$, and $D_n \subseteq \pi(B)$, $\forall n$. This entails $M(\bigcup D_n) = M(\bigcup \pi(B_n))$, the left-hand side equals $\sum_n M(D_n) = \sum_n M(\pi(B_n)) = M(\pi(B))$. Therefore

$$\sum_{k=1}^{\infty} \chi_{\pi(B_k)}(y) = \sum_{k=1}^{\infty} \chi_{D_k}(y) + \sum_{k=1}^{\infty} \chi_{F_k}(y),$$

where the second term on the right is equal to 0 a.e. M . Further,

$$\chi_{\pi(B)}(y) - \sum_{k=1}^n \chi_{\pi(B_k)}(y) = \chi_{\pi(B)}(y) - \sum_{k=1}^n \chi_{D_k} - \sum_{k=1}^n \chi_{F_k}(y) \geq 0 \text{ a.e. } M.$$

From

$$\int (\chi_{\pi(B)}(y) - \sum_{k=1}^n \chi_{\pi(B_k)}(y)) M(dy) \rightarrow 0,$$

we obtain, since the sub-integral function is bounded,

$$\int \lim_n (\chi_{\pi(B)}(y) - \sum_{k=1}^n \chi_{\pi(B_k)}(y)) M(dy) = 0,$$

which implies $\sum_{n=1}^{\infty} \chi_{\pi(B_n)}(y) = \chi_{\pi(B)}(y)$ a.e. M . This concludes the proof that $\nu(y, B) = \chi_{\pi(B)}(y)$ is a weak Markov kernel. \square

Our results so far can be summarized as follows.

THEOREM 3.5. *Let $M : (X, \mathcal{A}) \rightarrow \mathcal{E}(H)$ be a POV measure and $P : (Y, \mathcal{B}) \rightarrow \mathcal{E}(H)$ be a PV measure. The following conditions are equivalent:*

- (i) $\mathcal{R}(P) \subset \mathcal{R}(M)$,
- (ii) *there exists a weak Markov kernel ν with respect to M with values in $[0, 1]$, such that $P = \nu \circ M$,*
- (iii) *there exists a weak Markov kernel ν with respect to M , such that $P = \nu \circ M$.*

Moreover, if (Y, \mathcal{B}) is a standard Borel space, then the conditions are also equivalent to

- (ii') $P(B) = M(f^{-1}(B))$, $\forall B \in \mathcal{B}$, with $f : X \rightarrow Y$ measurable,
- (iii') *there is a Markov kernel λ , such that $P = \lambda \circ M$.*

We remark that the equivalence (i) \Leftrightarrow (ii') for real observables was proved in [8], where the proof was based on the Naimark theorem.

4. Clean observables

Let $M : (X, \mathcal{A}) \rightarrow \mathcal{E}(H)$ and $N : (Y, \mathcal{B}) \rightarrow \mathcal{E}(H)$ be two observables. We write $M \leq N$ if there exists a weak Markov kernel ν with respect to M , such that $N = \nu \circ M$. If also $N \leq M$, we write $M \sim N$. This defines an equivalence relation on the set of observables and \leq is a partial order on the equivalence classes. Minimal elements with respect to this order are called *clean*.

This equivalence and order have a statistical interpretation: if $M \leq N$, then the family of probability measures $\mathcal{P}_M := \{m \circ M : m \in \mathcal{S}\}$ is more informative than $\mathcal{P}_N := \{m \circ N : m \in \mathcal{S}\}$, in the sense that the elements of \mathcal{P}_M can be distinguished more precisely by statistical procedures than elements of \mathcal{P}_N , [20]. We remark that previous definitions of \leq involved smearings with respect to Markov kernels rather than weak Markov kernels. In the case of standard Borel spaces, the two notions are equivalent, whereas in the general situation, the weaker definition seems to be more appropriate.

The results of the previous sections can be applied to the characterization of cleanness of sharp observables. For this, we need the following simple observation.

LEMMA 4.1. *If a projection P is contained in the range of an observable M , then P commutes with all elements of $\mathcal{R}(M)$.*

COROLLARY 4.1. *A PV measure is clean iff its range generates a maximal abelian von Neumann subalgebra of $\mathcal{B}(H)$.*

Proof: Let E be a PV measure. Let the abelian von Neumann subalgebra \mathcal{N} generated by $\mathcal{R}(E)$ be not maximal. Then there is a maximal abelian von Neumann subalgebra \mathcal{M} which contains \mathcal{N} and a PV measure F , such that $\mathcal{R}(F)$ generates \mathcal{M} . Then $\mathcal{R}(E) \subset \mathcal{R}(F)$. By Theorem 3.5, E is a smearing of F , but F is not a smearing of E . Therefore E is not clean.

Assume that $\mathcal{R}(E)$ generates a maximal abelian von Neumann subalgebra \mathcal{M} . Since H is separable, there is a self-adjoint operator T such that $\mathcal{R}(T) = \mathcal{R}(E)$, and $\mathcal{M} = \{T\}''$. In particular, every projection P in \mathcal{M} belongs to $\mathcal{R}(T) = \mathcal{R}(E)$.

Suppose that E is a smearing of a POV measure M . Then $\mathcal{R}(E) \subset \mathcal{R}(M)$, and $E(B)M(C) = M(C)E(B)$ for all B, C by Lemma 4.1. Therefore $\mathcal{R}(M) \subset \mathcal{M}' = \mathcal{M} = \{T\}''$. As in the proof of Theorem 3.1, this implies that M is a smearing of E . \square

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