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# Atomic Intersection of s-Fields and Some of Its Consequences

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### Quaderni di Dipartimento

## Atomic Intersection of $\sigma$ -Fields and Some of Its Consequences

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### ATOMIC INTERSECTION OF $\sigma$ -FIELDS AND SOME OF ITS CONSEQUENCES

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ABSTRACT. Let  $(\Omega, \mathcal{F}, P)$  be a probability space. For each  $\mathcal{G} \subset \mathcal{F}$ , define  $\overline{\mathcal{G}}$  as the  $\sigma$ -field generated by  $\mathcal{G}$  and those sets  $F \in \mathcal{F}$  satisfying  $P(F) \in \{0, 1\}$ . Conditions for P to be atomic on  $\cap_{i=1}^k \overline{\mathcal{A}_i}$ , with  $\mathcal{A}_1, \ldots, \mathcal{A}_k \subset \mathcal{F}$  sub- $\sigma$ -fields, are given. Conditions for P to be 0-1-valued on  $\cap_{i=1}^k \overline{\mathcal{A}_i}$  are given as well. These conditions are useful in various fields, including Gibbs sampling, iterated conditional expectations and the intersection property.

#### 1. Introduction

Throughout,  $(\Omega, \mathcal{F}, P)$  is a probability space,  $\mathcal{A}_1, \dots, \mathcal{A}_k \subset \mathcal{F}$  sub- $\sigma$ -fields,  $k \geq 2$ , and we let

$$\mathcal{N} = \{ F \in \mathcal{F} : P(F) \in \{0,1\} \}$$
 and  $\overline{\mathcal{G}} = \sigma(\mathcal{G} \cup \mathcal{N})$  for any  $\mathcal{G} \subset \mathcal{F}$ .

As discussed in Section 3, the sub- $\sigma$ -field

$$\mathcal{D} = \bigcap_{i=1}^k \overline{\mathcal{A}_i}$$

plays a role in various subjects, including Gibbs sampling, iterated conditional expectations and the intersection property. In a previous paper, in a Gibbs sampling framework, we investigated when  $\bigcap_{i=1}^k \overline{\mathcal{A}_i} = \overline{\bigcap_{i=1}^k \mathcal{A}_i}$ ; see [3].

In this paper, instead, we focus on atomicity of P on  $\mathcal{D}$ . In fact, atomicity of  $P|\mathcal{D}$  (i.e., the restriction of P to  $\mathcal{D}$ ) has implications in each of the subjects mentioned above. It turns out that  $P|\mathcal{D}$  is actually atomic under mild conditions.

An extreme form of atomicity for  $P|\mathcal{D}$  is  $\mathcal{D} = \mathcal{N}$ , that is, P 0-1-valued on  $\mathcal{D}$ . Indeed,  $\mathcal{D} = \mathcal{N}$  is fundamental for Gibbs sampling and very useful for the intersection property; see [3], [7] and [10].

Our main results are in Sections 4 and 5. Section 4 gives general results on atomicity of  $P|\mathcal{D}$ . It includes a characterization (Theorem 2), a criterion for identifying the atoms (Theorem 3) and a sufficient condition (Theorem 4). Section 5, motivated by Gibbs sampling applications, concerns the particular case

$$\mathcal{A}_i = \sigma(X_1, \dots, X_{i-1}, X_{i+1}, \dots, X_k)$$

where  $X_1, \ldots, X_k$  are any random variables on  $(\Omega, \mathcal{F}, P)$ . It contains working sufficient conditions for  $\mathcal{D} = \mathcal{N}$  (Theorem 8) and for  $P|\mathcal{D}$  to be atomic (Theorem 10). Indeed,  $P|\mathcal{D}$  is atomic whenever the probability distribution of  $(X_1, \ldots, X_k)$  is absolutely continuous with respect to a  $\sigma$ -finite product measure.

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Finally, it is worth noting that  $\mathcal{D} = \mathcal{N}$  whenever  $\mathcal{A}_r$  is independent of  $\mathcal{A}_s$  for some r, s. Given  $D \in \mathcal{D}$ , in fact, one has  $P(A_i \Delta D) = 0$  for some  $A_i \in \mathcal{A}_i$ ,  $i = 1, \ldots, k$ . Hence,  $P(D) = P(A_r \cap A_s) = P(A_r)P(A_s) = P(D)^2$ .

### 2. Preliminaries

Let  $(\mathcal{X}, \mathcal{E}, Q)$  be a probability space. A Q-atom is any set  $H \in \mathcal{E}$  such that Q(H) > 0 and  $Q(\cdot \mid H)$  is 0-1-valued. In general, there are three possible situations: (i) Q is nonatomic, i.e., there are no Q-atoms; (ii) Q is atomic, i.e., the Q-atoms form a (countable) partition of  $\mathcal{X}$ ; (iii) there is  $K \in \mathcal{E}$ , 0 < Q(K) < 1, such that  $Q(\cdot \mid K)$  is nonatomic and  $K^c$  is a (countable) disjoint union of Q-atoms.

Thus,  $D \subset \Omega$  is an atom of  $P|\mathcal{D}$  if and only if  $D \in \mathcal{D}$ , P(D) > 0 and  $P(\cdot \mid D)$  is 0-1-valued on  $\mathcal{D}$ . In the sequel, when  $P|\mathcal{D}$  is atomic, we also say that  $\mathcal{D}$  is atomic under P.

For later purposes, we also note that Q is nonatomic if and only if  $(\mathcal{X}, \mathcal{E}, Q)$  supports a real random variable with uniform distribution on (0,1). In fact, if U is a uniform random variable on  $(\mathcal{X}, \mathcal{E}, Q)$ , then Q is nonatomic since  $\sigma(U) \subset \mathcal{E}$  and  $Q|\sigma(U)$  is nonatomic. Conversely, by Lyapunov's convexity theorem, if Q is nonatomic the range of Q is [0,1]; see e.g. [8] or Theorem 5.1.6 of [4] for a proof (based on transfinite induction or Zorn's lemma, respectively). Since the range of Q is [0,1], a uniform random variable on  $(\mathcal{X},\mathcal{E},Q)$  can be obtained by arguing as in the proof of Lemma 2 of [1]; see also Theorem 3.1 of [2].

We finally recall that, for any sub- $\sigma$ -field  $\mathcal{G} \subset \mathcal{F}$ ,

$$\overline{\mathcal{G}} = \{ F \in \mathcal{F} : P(F\Delta G) = 0 \text{ for some } G \in \mathcal{G} \}.$$

A straightforward consequence is that a real  $\overline{\mathcal{G}}$ -measurable function on  $\Omega$  coincides a.s. with some  $\mathcal{G}$ -measurable function. Thus, if  $U:\Omega\to\mathbb{R}$  is  $\mathcal{D}$ -measurable, then  $U=U_i$  a.s. for some  $\mathcal{A}_i$ -measurable function  $U_i:\Omega\to\mathbb{R},\ i=1,\ldots,k$ .

### 3. Fields where $\mathcal{D}$ appears

We list some fields involving  $\mathcal{D}$ , by paying particular attention to the case where  $P|\mathcal{D}$  is atomic. We stress by now that, for atomicity of  $P|\mathcal{D}$  to be a real advantage, the atoms of  $P|\mathcal{D}$  and their probabilities should be known.

Throughout,  $X_i$  is a random variable on  $(\Omega, \mathcal{F}, P)$  with values in the measurable space  $(\mathcal{X}_i, \mathcal{B}_i)$ , i = 1, ..., k, and

$$X_i^* = (X_1, \dots, X_{i-1}, X_{i+1}, \dots, X_k).$$

3.1. **Intersection property.** Let X be a random variable on  $(\Omega, \mathcal{F}, P)$  with values in the measurable space  $(\mathcal{X}, \mathcal{E})$ . The intersection property (IP) is

$$X \perp X_i^* \mid X_i \text{ for } i = 1, \dots, k \implies X \perp (X_1, \dots, X_k)$$

where the notation " $U \perp V \mid W$ " stands for "U conditionally independent of V given W". It is well known that IP may fail. As a trivial example, take X not independent of  $X_1$  and  $X_i = X_1$  for all i.

IP is involved in a number of arguments. It appears, for instance, in graphical models, zero entries in contingency tables, causal inference and estimation in Markov processes; see [10] and references therein.

The connections between IP and  $\mathcal{D}$  are made clear by part (b) of the next (obvious) result. Part (a) is already known for k=2 (see Proposition 2.2 of [10] and references therein) but we give a proof to make the paper self-contained.

**Theorem 1.** Let  $A_i = \sigma(X_i)$  for all i. Then:

- (a)  $X \perp X_i^* \mid X_i \text{ for } i = 1, ..., k \iff E(f(X) \mid X_1, ..., X_k) = E(f(X) \mid \mathcal{D}) \text{ a.s.}$  for each real bounded measurable function f on  $(\mathcal{X}, \mathcal{E})$ ;
- **(b)**  $X \perp X_i^* \mid X_i \text{ for } i = 1, \ldots, k \text{ and } X \perp \mathcal{D} \iff X \perp (X_1, \ldots, X_k);$
- (c)  $X \perp D$  if and only if

$$P(X \in A, X_1 \in B_1) = P(X \in A) P(X_1 \in B_1)$$
 whenever

$$A \in \mathcal{E}, B_1 \in \mathcal{B}_1 \text{ and } P(\{X_1 \in B_1\} \Delta \{X_i \in B_i\}) = 0 \text{ for some } B_i \in \mathcal{B}_i, i = 2, \dots, k.$$

*Proof.* (a) Suppose  $E(f(X) \mid X_1, ..., X_k) = E(f(X) \mid \mathcal{D})$  a.s. for all bounded measurable f on  $(\mathcal{X}, \mathcal{E})$ . Given i, since  $\mathcal{D} \subset \overline{\sigma(X_i)} \subset \overline{\sigma(X_1, ..., X_k)}$ , then

$$E(f(X) | X_i) = E(f(X) | D) = E(f(X) | X_1, ..., X_k)$$
 a.s.

for all bounded measurable f, that is,  $X \perp X_i^* \mid X_i$ . Conversely, suppose  $X \perp X_i^* \mid X_i$  for all i. Given a bounded measurable f,

$$E(f(X) | X_1) = E(f(X) | X_1, \dots, X_k) = E(f(X) | X_i)$$
 a.s. for all *i*.

Thus,  $E(f(X) \mid X_1)$  is  $\overline{\sigma(X_i)}$ -measurable for all i, i.e., it is  $\mathcal{D}$ -measurable. Therefore,  $E(f(X) \mid X_1, \dots, X_k) = E(f(X) \mid X_1) = E(f(X) \mid \mathcal{D})$  a.s..

- (b) "\(\iff \)" is obvious. As to "\(\iff \)", it suffices noting that  $E(f(X) \mid X_1, \dots, X_k) = E(f(X) \mid \mathcal{D}) = Ef(X)$  a.s. for all bounded measurable f, where the first equality is by part (a) and the second is because  $X \perp \mathcal{D}$ .
  - (c) Just note that

$$\mathcal{D} = \{ F \in \mathcal{F} : P(F\Delta\{X_i \in B_i\}) = 0 \text{ for some } B_i \in \mathcal{B}_i, \ i = 1, \dots, k \}.$$

Thus,  $X \perp \mathcal{D}$  is a (natural) sufficient condition for IP. In a sense, it is necessary as well, since it is a consequence of  $X \perp (X_1, \ldots, X_k)$ . Heuristically,  $X \perp \mathcal{D}$  means that X is not affected by that part of information which is common to  $X_1, \ldots, X_k$ .

To test whether  $X \perp \mathcal{D}$ , atomicity of  $\mathcal{D}$  under P can help. If  $P \mid \mathcal{D}$  is atomic, in fact,  $X \perp \mathcal{D}$  reduces to

$$P(X \in A, D) = P(X \in A) P(D)$$
 for all  $A \in \mathcal{E}$  and atoms  $D$  of  $P|\mathcal{D}$ .

As shown in Theorem 4, for  $P|\mathcal{D}$  to be atomic, it is enough that the distribution of  $(X_1, \ldots, X_k)$  is absolutely continuous with respect to a  $\sigma$ -finite product measure; see also Lemma 6.

A last note is that  $X \perp \mathcal{D}$  is trivially true whenever  $\mathcal{D} = \mathcal{N}$ . In [10], a paper which inspired the present Subsection,  $\mathcal{D} = \mathcal{N}$  was firstly viewed as a sufficient condition for IP. In [3], in a Gibbs sampling framework,  $\mathcal{D} = \mathcal{N}$  was given a characterization and various sufficient conditions.

3.2. Iterated conditional expectations. Let X be a real random variable on  $(\Omega, \mathcal{F}, P)$  such that  $EX^2 < \infty$  and

$$\mathcal{D}_{mk+i} = \mathcal{A}_i$$
 for all  $m = 0, 1, \dots$  and  $i = 1, \dots, k$ .

Define  $Z_0 = X$  and  $Z_n = E(Z_{n-1} \mid \mathcal{D}_n)$  for  $n \ge 1$ . Then,

(1) 
$$Z_n \stackrel{a.s.}{\to} E(X \mid \mathcal{D}) \text{ as } n \to \infty.$$

This classical result was obtained by Burkholder-Chow [5] for k=2 and Delyon-Delyon [6] for arbitrary k. See [7] for some historical notes.

Suppose now that  $\mathcal{D}$  is atomic under P and the goal is estimating EX. Then,  $E(X \mid \mathcal{D}) = \sum_{j} I_{D_j} E(X \mid D_j)$  a.s. where  $D_1, D_2, \ldots$  denote the (disjoint) atoms of  $P \mid \mathcal{D}$ . Thus, one should apply relation (1) on each atom  $D_j$ , so as to obtain an estimate for  $E(X \mid D_j)$ , and then use the formula  $EX = \sum_{j} P(D_j) E(X \mid D_j)$ .

3.3. **Gibbs sampling.** As noted in [7], the limit theorem of Burkholder-Chow and Delyon-Delyon (Subsection 3.2) is intrinsically connected to Gibbs sampling.

Let  $X_1, \ldots, X_k$  be the canonical projections on

$$(\Omega, \mathcal{F}) = (\prod_{i=1}^k \mathcal{X}_i, \prod_{i=1}^k \mathcal{B}_i).$$

Each  $X_i$  is assumed to admit a regular conditional distribution  $\gamma_i$  given  $X_i^*$ . In the notation  $u=(x_1,\ldots,x_{i-1},x_{i+1},\ldots,x_k)$ , where  $x_j\in\mathcal{X}_j$  for all  $j\neq i$ , this means that: (i)  $\gamma_i(u)$  is a probability measure on  $\mathcal{B}_i$  for every u; (ii)  $u\mapsto\gamma_i(u)(A)$  is measurable for  $A\in\mathcal{B}_i$ ; (iii)  $P(B)=\int\int I_B(x_1,\ldots,x_k)\gamma_i(u)(dx_i)\,\gamma_i^*(du)$  for  $B\in\prod_{i=1}^k\mathcal{B}_i$ , where  $\gamma_i^*$  denotes the marginal distribution of  $X_i^*$ .

The Gibbs-chain

$$Y_n = (Y_{1,n}, \dots, Y_{k,n}), \quad n \ge 0,$$

can be informally described as follows. Starting from  $\omega = (x_1, \ldots, x_k)$ , the next state  $\omega^* = (a_1, \ldots, a_k)$  is obtained by sequentially generating  $a_k, a_{k-1}, \ldots, a_1$ , each  $a_i$  being selected from the conditional distribution of  $X_i$  given  $X_1 = x_1, \ldots, X_{i-1} = x_{i-1}, X_{i+1} = a_{i+1}, \ldots, X_k = a_k$ . Formally,  $(Y_n)$  is the homogeneous Markov chain with state space  $(\Omega, \mathcal{F})$  and transition kernel

$$K(\omega, B) = K((x_1, \dots, x_k), B)$$

$$= \int \dots \int I_B(a_1, \dots, a_k) \prod_{i=1}^k \gamma_i(x_1, \dots, x_{i-1}, a_{i+1}, \dots, a_k) (da_i).$$

Note that P is a stationary distribution for  $(Y_n)$ , i.e.,  $P(\cdot) = \int K(\omega, \cdot) P(d\omega)$ . Let  $\mathbb{P}$  denote the law of  $(Y_n)$  such that  $Y_0 \sim P$ .

The Gibbs chain is constructed mainly for sampling from P. To this end, the following SLLN is fundamental

(2) 
$$\frac{1}{n} \sum_{i=0}^{n-1} \phi(Y_i) \to \int \phi \, dP, \ \mathbb{P}\text{-a.s., for all } \phi \in L_1(P).$$

Another basic requirement of  $(Y_n)$ , stronger than (2), is ergodicity on some set  $S \in \mathcal{F}$ , that is

$$P(S)=1, \quad K(\omega,S)=1 \quad \text{and} \quad \|K^n(\omega,\cdot)-P\| \to 0 \text{ for each } \omega \in S,$$

where  $\|\cdot\|$  is total variation norm and  $K^n$  the *n*-th iterate of K.

Now, letting  $A_i = \sigma(X_i^*)$  for all i, the SLLN under (2) is equivalent to

$$\mathcal{D} = \mathcal{N}$$
.

In addition, in case  $\mathcal{F}$  is countably generated and P absolutely continuous with respect to a  $\sigma$ -finite product measure,  $\mathcal{D} = \mathcal{N}$  if and only if  $(Y_n)$  is ergodic on  $S_0 = \{\omega \in \Omega : K(\omega, \cdot) \ll P\}$ . See Theorems 4.2, 4.5 and Remark 4.7 of [3].

Conditions for  $\mathcal{D} = \mathcal{N}$  (when  $\mathcal{A}_i = \sigma(X_i^*)$  for all i) are given in Theorem 8; see also Lemma 6.

Strictly speaking, thus, Gibbs sampling is admissible only if  $\mathcal{D} = \mathcal{N}$ . At least in principle, however, it makes sense even if  $\mathcal{D} \neq \mathcal{N}$ , provided  $\mathcal{D}$  is atomic under P. In fact, if  $P|\mathcal{D}$  is atomic (with disjoint atoms  $D_1, D_2, \ldots$ ), then (2) turns into

(2\*) 
$$\frac{1}{n} \sum_{i=0}^{n-1} \phi(Y_i) \to \int \phi(\omega) P(d\omega \mid D_j),$$

$$\mathbb{P}\text{-a.s. on } \{Y_0 \in D_j\}, \text{ for all } j \text{ and } \phi \in L_1(P).$$

The SLLN under  $(2^*)$  can be proved by the same argument used for Theorem 4.2 of [3] plus the observation that  $K(\cdot, B) = I_B(\cdot)$ , P-a.s., for each  $B \in \mathcal{D}$ .

If the atoms  $D_j$  are only a finite number, and they are known together with their probabilities  $P(D_j)$ , then  $(2^*)$  can be used to evaluate  $\int \phi \, dP$ . The chain  $(Y_n)$  should be started on each  $D_j$ , so as to obtain an estimate for  $\int \phi(\omega) \, P(d\omega \mid D_j)$ , and then the formula  $\int \phi \, dP = \sum_j P(D_j) \int \phi(\omega) \, P(d\omega \mid D_j)$  should be applied.

As shown in Theorem 10, for  $P|\mathcal{D}$  to be atomic, it is enough that the distribution of  $(X_1, \ldots, X_k)$  is absolutely continuous with respect to a  $\sigma$ -finite product measure.

### 4. Atomicity of $\mathcal{D}$ under P

We begin with a definition. Say that  $H \subset \Omega$  has the trivial intersection property, or briefly that H is TIP, in case  $H \in \mathcal{F}$ , P(H) > 0, and

$$A_i \in \mathcal{A}_i$$
 and  $P(A_i \Delta A_1 \mid H) = 0$  for  $i = 1, \dots, k \Longrightarrow P(A_1 \mid H) \in \{0, 1\}.$ 

Here are some obvious consequences of the definition.

- (i) If H is TIP,  $A_i \in \mathcal{A}_i$  and  $P(A_i \Delta A_1 \mid H) = 0$  for all i, then either  $P(A_i \mid H) = 0$  for all i or  $P(A_i \mid H) = 1$  for all i.
- (ii) Let  $H \in \mathcal{F}$  with P(H) > 0 and write  $P_H = P(\cdot \mid H)$ . Then, H is TIP if and only if  $\Omega$  is  $P_H$ -TIP (i.e.,  $\Omega$  is TIP under  $P_H$ ). Moreover,  $\Omega$  is TIP if and only if  $\mathcal{D} = \mathcal{N}$ . Therefore, the definition of TIP set may be rephrased as follows: H is TIP if and only if

$$\mathcal{D}_{P_H} = \mathcal{N}_{P_H}$$
 where  $\mathcal{N}_{P_H} = \{F \in \mathcal{F} : P_H(F) \in \{0,1\}\}$  and  $\mathcal{D}_{P_H} = \cap_{i=1}^k \sigma(\mathcal{A}_i \cup \mathcal{N}_{P_H})$ .

(iii) Let Q be a probability on  $\mathcal{F}$ . If P and Q are equivalent (i.e.,  $P \ll Q$  and  $Q \ll P$ ), then H is Q-TIP if and only if it is P-TIP. If  $P \ll Q$  and  $H \subset \{\frac{dP}{dQ} > 0\}$ , for some given version of  $\frac{dP}{dQ}$ , then H is Q-TIP if and only if it is P-TIP.

The present notion of TIP set generalizes the one given in [3] for k = 2. Among other things, such a notion is basic for characterizing atomicity of  $P|\mathcal{D}$ .

**Theorem 2.** Let  $H \subset \Omega$ . Then,

- (a) If H is TIP, there is an atom  $H^*$  of  $P|\mathcal{D}$  satisfying  $H^* \supset H$  and  $P(D \mid H^*) = P(D \mid H)$  for all  $D \in \mathcal{D}$ ;
- (b) For H to be an atom of  $P|\mathcal{D}$  it is necessary and sufficient that  $H \in \mathcal{D}$  and H is TIP;
- (c)  $\mathcal{D}$  is atomic under P if and only if  $P(\cup_n H_n) = 1$  for some countable collection  $H_1, H_2, \ldots$  of TIP sets.

Proof. (a) Suppose H is TIP. We first prove that  $P(\cdot \mid H)$  is 0-1 on  $\mathcal{D}$ . Given  $D \in \mathcal{D}$ , for each i there is  $A_i \in \mathcal{A}_i$  such that  $P(A_i \Delta D) = 0$ . Hence,  $P((A_i \Delta A_1) \cap H) \leq P(A_i \Delta A_1) = 0$  for all i, and H TIP implies  $P(D \mid H) = P(A_1 \mid H) \in \{0, 1\}$ . Next, by a standard argument, there is  $H^* \in \mathcal{D}$  such that  $H^* \supset H$  and

$$P(H^*) = \inf\{P(D) : H \subset D \in \mathcal{D}\}.$$

Let  $D \in \mathcal{D}$ . If  $P(D \mid H) = 1$ , then

$$H \subset (D \cap H^*) \cup (D^c \cap H) \in \mathcal{D},$$

so that  $P(H^*) \leq P((D \cap H^*) \cup (D^c \cap H)) = P(D \cap H^*)$  by definition of  $H^*$ . Hence,  $P(D \mid H^*) = 1$ . Taking complements, if  $P(D \mid H) = 0$  then  $P(D \mid H^*) = 0$ . Thus,  $H^*$  is an atom of  $P|\mathcal{D}$  and  $P(\cdot \mid H^*) = P(\cdot \mid H)$  on  $\mathcal{D}$ .

(b) If  $H \in \mathcal{D}$  is TIP, then H is an atom of  $P|\mathcal{D}$  since  $P(H \mid H^*) = P(H \mid H) = 1$ , where  $H^*$  is as in point (a). Conversely, suppose H is an atom of  $P|\mathcal{D}$ . To prove H TIP, we fix  $A_i \in \mathcal{A}_i$  such that  $P(A_i \Delta A_1 \mid H) = 0$  for  $i = 1, \ldots, k$ . For each i, since  $H \in \mathcal{D} \subset \overline{\mathcal{A}_i}$ , some  $H_i \in \mathcal{A}_i$  meets  $P(H \Delta H_i) = 0$ . Moreover,

$$P((A_1 \cap H)\Delta(A_i \cap H_i)) \le P(H\Delta H_i) + P((A_i\Delta A_1) \cap H) = 0.$$

Hence,  $A_1 \cap H \in \overline{A_i}$  for all i, that is,  $A_1 \cap H \in \mathcal{D}$ . Since H is an atom of  $P|\mathcal{D}$ , it follows that  $P(A_1 \mid H) = P(A_1 \cap H \mid H) \in \{0, 1\}$ . Thus, H is TIP.

(c) If  $P|\mathcal{D}$  is atomic, it suffices to take the  $H_n$  as the atoms of  $P|\mathcal{D}$  and to apply point (b). Conversely, if  $P(\cup_n H_n) = 1$  with the  $H_n$  TIP, for each n point (a) implies  $H_n \subset H_n^*$  for some atom  $H_n^*$  of  $P|\mathcal{D}$ . Then,  $P|\mathcal{D}$  is atomic since  $P(\cup_n H_n^*) \geq P(\cup_n H_n) = 1$ .

By Theorem 2,  $P|\mathcal{D}$  is atomic provided  $\Omega$  can be covered by countably many TIP sets  $H_1, H_2, \ldots$  In this case, every atom D admits the representation  $D = \bigcup_{i \in I} H_i$  a.s. for some index set I (by point (a)). The next issue, thus, is identifying such atoms using the  $H_n$  as building blocks. Indeed, the atoms are maximal TIP sets, according to the following result.

**Theorem 3.** Suppose  $P(\cup_n H_n) = 1$ , where  $H_1, H_2, \ldots$  are TIP, and let  $D \subset \Omega$ . Then, D is an atom of  $P|\mathcal{D}$  if and only if D is TIP and

(3) 
$$D \cup H_n$$
 fails to be TIP whenever  $P(H_n \setminus D) > 0$ .

*Proof.* By Theorem 2, it can be assumed D TIP, and we have to prove that condition (3) is equivalent to  $D \in \mathcal{D}$ . Suppose (3) holds. Let  $N = \{n : P(H_n \setminus D) > 0\}$ . If  $N = \emptyset$ , then  $P(D^c) \leq \sum_n P(H_n \setminus D) = 0$ , so that  $D \in \mathcal{N} \subset \mathcal{D}$ . If  $N \neq \emptyset$ , by (3), for each  $n \in N$  there are  $A_{i,n} \in \mathcal{A}_i$ ,  $i = 1, \ldots, k$ , satisfying

$$P(A_{i,n}\Delta A_{1,n}\mid D\cup H_n)=0$$
 and  $P(A_{i,n}\mid D\cup H_n)\in (0,1)$  for all  $i$ .

Since D and  $H_n$  are TIP, one also has  $P(A_{i,n} \mid D) \in \{0,1\}$  and  $P(A_{i,n} \mid H_n) \in \{0,1\}$  for all i, and thus

$$P(A_{i,n} | D) = 1 - P(A_{i,n} | H_n)$$
 for all i.

Define  $F_{i,n} = A_{i,n}$  or  $F_{i,n} = A_{i,n}^c$  as  $P(A_{i,n} \mid D) = 1$  or  $P(A_{i,n} \mid D) = 0$ , and  $A_i = \bigcap_{n \in N} F_{i,n}$ .

Then,  $P(A_i \mid D) = 1$  and  $P(A_i \mid H_n) = 0$  for all i and  $n \in N$ . Hence, given i,

$$P(A_i \Delta D) = P(A_i \setminus D) \le \sum_{n \in N} P(A_i \cap D^c \cap H_n) \le \sum_{n \in N} P(A_i \cap H_n) = 0.$$

Since  $A_i \in \mathcal{A}_i$ , it follows that  $D \in \overline{\mathcal{A}_i}$ , that is,  $D \in \mathcal{D}$ . Conversely, suppose  $D \in \mathcal{D}$  and  $D \cup H_n$  is TIP for some n. Since  $P(D \mid D \cup H_n) \in \{0,1\}$  (by point (a) of Theorem 2) and P(D) > 0 (as D is TIP), then  $P(H_n \setminus D) = 0$ .

In real problems, it is not unusual that  $P \ll Q$  for some probability Q on  $\mathcal{F}$  which makes  $\mathcal{A}_1, \ldots, \mathcal{A}_k$  independent. This does not imply  $\mathcal{D} = \mathcal{N}$  (see Examples 3.16 and 3.17 of [3]) but it suffices for atomicity of  $P|\mathcal{D}$ . Actually, it is enough that a couple of the  $\mathcal{A}_i$  are independent under Q.

**Theorem 4.**  $P|\mathcal{D}$  is atomic provided  $P \ll Q$  for some probability measure Q on  $\mathcal{F}$  which makes  $\mathcal{A}_r$  and  $\mathcal{A}_s$  independent for some r, s.

Proof. Fix  $H \in \mathcal{D}$  with P(H) > 0 and let  $P_H = P(\cdot \mid H)$ . If  $P_H \mid \mathcal{D}$  is nonatomic, the probability space  $(\Omega, \mathcal{D}, P_H \mid \mathcal{D})$  supports a real random variable with uniform distribution; see Section 2. Hence, it suffices to prove that each  $\mathcal{D}$ -measurable function  $U: \Omega \to \mathbb{R}$  satisfies  $P_H(U \in C) = 1$  for some countable set  $C \subset \mathbb{R}$ . Further, since  $P_H \ll P$ , it is enough to show that  $P(U \in C) = 1$ . Let  $U: \Omega \to \mathbb{R}$  be  $\mathcal{D}$ -measurable. Then,  $U = U_i$  a.s. for some  $U_i: \Omega \to \mathbb{R}$  satisfying  $\sigma(U_i) \subset \mathcal{A}_i$ ,  $i = 1, \ldots, k$ . Define the countable set  $C = \{c \in \mathbb{R} : Q(U_s = c) > 0\}$ . Since  $U_r$  and  $U_s$  are independent under Q,

$$Q(U_r \notin C, U_r = U_s) = \int_{\{U_r \notin C\}} Q\{x : U_s(x) = U_r(\omega)\} Q(d\omega) = 0.$$

Thus,  $P \ll Q$  yields

$$P(U \in C) = 1 - P(U \notin C, U = U_r = U_s) = 1 - P(U_r \notin C, U_r = U_s) = 1.$$

For k=2, Theorem 4 reduces to Theorem 3.10 of [3].

**Remark 5.** Let  $A_i = \sigma(X_i)$  for all i, where  $X_i : \Omega \to \mathcal{X}_i$  is a random variable and  $\mathcal{X}_i$  a separable metric space (equipped with its Borel  $\sigma$ -field  $\mathcal{B}_i$ ). Then,  $P|\mathcal{D}$  need not be atomic even though

$$(4) P(X_i = f(X_i)) = 0$$

for all  $i \neq j$  and all measurable functions  $f: \mathcal{X}_j \to \mathcal{X}_i$ . We mention this fact since, for some time, we conjectured  $P|\mathcal{D}$  atomic under (4).

As an example, let k=2,  $X_1=(U,W)$  and  $X_2=(V,W)$ , where U,V,W are real independent random variables with nonatomic distributions. Take  $\mathcal{X}_i=\mathbb{R}^2$  and  $\mathcal{A}_i=\sigma(X_i)$  for i=1,2. Given a measurable function  $f=(f_1,f_2):\mathbb{R}^2\to\mathbb{R}^2$ , one obtains  $P(X_2=f(X_1))\leq P(V=f_1(U,W))=0$  since V has nonatomic distribution and is independent of (U,W). Likewise,  $P(X_1=f(X_2))=0$ . However,  $P|\mathcal{D}$  is nonatomic, as  $\sigma(W)\subset\mathcal{D}$  and W has nonatomic distribution.

Finally, we state a simple but useful fact as a lemma. Let Q be a probability measure on  $\mathcal{F}$ . Say that P and Q are equivalent on rectangles in case

$$P(A) = 0 \Leftrightarrow Q(A) = 0 \text{ for each } A \in \mathcal{R},$$
  
where  $\mathcal{R} = \{ \cap_{i=1}^k A_i : A_i \in \mathcal{A}_i, i = 1, \dots, k \}.$ 

If  $A \in \mathcal{R}$ , then  $A^c$  is a finite union of elements of  $\mathcal{R}$ . Hence,  $P(A) = 1 \Leftrightarrow Q(A) = 1$  and  $P(A\Delta B) = 0 \Leftrightarrow Q(A\Delta B) = 0$  whenever  $A, B \in \mathcal{R}$  and  $P(A\Delta B) = 0$  are equivalent on rectangles. Note that  $A_i \subset \mathcal{R}$  for all i. Note also that P and Q need not be equivalent on  $\sigma(\mathcal{R})$  even though they are equivalent on rectangles.

**Lemma 6.** Let P and Q be equivalent on rectangles. If D is an atom of  $Q|\mathcal{D}_Q$ , there is  $A \in \mathcal{R}$  such that  $Q(A\Delta D) = 0$  and A is an atom of  $P|\mathcal{D}_P$ . Moreover,

$$\mathcal{D}_Q = \mathcal{N}_Q \Leftrightarrow \mathcal{D}_P = \mathcal{N}_P, \ and$$

 $\mathcal{D}_Q$  is atomic under  $Q \Leftrightarrow \mathcal{D}_P$  is atomic under P.

(Here,  $\mathcal{N}_Q = \{ F \in \mathcal{F} : Q(F) \in \{0,1\} \}$ ,  $\mathcal{D}_Q = \bigcap_{i=1}^k \sigma(\mathcal{A}_i \cup \mathcal{N}_Q)$ ,  $\mathcal{N}_P = \mathcal{N}$  and  $\mathcal{D}_P = \mathcal{D}$ ).

Proof. We first prove that, for each  $D \in \mathcal{D}_Q$  with Q(D) > 0, there is A = A(D) satisfying  $A \in \mathcal{A}_1 \cap \mathcal{D}_P$ ,  $Q(A \Delta D) = 0$  and P(A) > 0. Take in fact  $A_i \in \mathcal{A}_i$  with  $Q(A_i \Delta D) = 0$ ,  $i = 1, \ldots, k$ , and let  $A = A_1$ . Then  $A \in \mathcal{A}_1$ ,  $Q(A \Delta D) = 0$  and P(A) > 0. Since  $Q(A_i \Delta A) = 0$  for all i, then  $P(A_i \Delta A) = 0$  for all i, so that  $A \in \mathcal{D}_P$ . Next, let D be an atom of  $Q|\mathcal{D}_Q$  and A = A(D). Then,  $A \in \mathcal{A}_1 \subset \mathcal{R}$ . Given  $G \in \mathcal{D}_P$ , for each i there is  $G_i \in \mathcal{A}_i$  such that  $P(G \Delta G_i) = 0$ . Again,  $P(G_i \Delta G_1) = 0$  for all i implies  $Q(G_i \Delta G_1) = 0$  for all i, so that  $G_1 \in \mathcal{D}_Q$ . Since A is an atom of  $Q|\mathcal{D}_Q$  (as  $Q(A \Delta D) = 0$ ), either  $Q(A \cap G_1) = 0$  or  $Q(A \cap G_1^c) = 0$ . Accordingly, either  $P(A \cap G) = P(A \cap G_1) = 0$  or  $P(A \cap G_1) = 0$  or  $P(A \cap G_1) = 0$ , i.e., A is an atom of  $P|\mathcal{D}_P$ . Next, if  $\mathcal{D}_Q = \mathcal{N}_Q$ , then  $\Omega$  is an atom of  $Q|\mathcal{D}_Q$ . Thus,  $A = A(\Omega)$  is an atom of  $P|\mathcal{D}_P$  and P(A) = 1, i.e.,  $\mathcal{D}_P = \mathcal{N}_P$ . Finally, suppose  $Q|\mathcal{D}_Q$  is atomic with (disjoint) atoms  $D_1, D_2, \ldots$ . Let  $A_j = A(D_j)$  and  $A = \cup_j A_j$ . Then, each  $A_j$  is an atom of  $P|\mathcal{D}_P$ , and P(A) = 1 since Q(A) = 1 and Q(A)

### 5. Applications to Gibbs sampling

As remarked in Subsection 3.3, in a Gibbs sampling framework it is fundamental that  $\mathcal{D} = \mathcal{N}$ , or at least that  $\mathcal{D}$  is atomic under P, when the  $\mathcal{A}_i$  are given by

$$\mathcal{A}_i = \sigma(X_1, \dots, X_{i-1}, X_{i+1}, \dots, X_k).$$

In this section, we let  $\mathcal{A}_i = \sigma(X_i^*)$  for all i, where  $X_i^* = (X_1, \dots, X_{i-1}, X_{i+1}, \dots, X_k)$  and the  $X_i$  are random variables on  $(\Omega, \mathcal{F}, P)$  with values in the measurable spaces  $(\mathcal{X}_i, \mathcal{B}_i)$ ,  $i = 1, \dots, k$ . We also let  $\mathcal{D}_0 = \cap_i \overline{\sigma(X_i)}$ . Since  $\mathcal{D}_0 \subset \mathcal{D}$ , P is 0-1-valued or atomic on  $\mathcal{D}_0$  whenever it is so on  $\mathcal{D}$ .

Let  $\mathcal{X} = \prod_{i=1}^k \mathcal{X}_i$  and let  $\mathcal{B} = \prod_{i=1}^k \mathcal{B}_i$  denote the product  $\sigma$ -field on  $\mathcal{X}$ . Define two measures on  $\mathcal{B}$  as

$$\gamma(\cdot) = P((X_1, \dots, X_k) \in \cdot)$$
 and  $\mu = \mu_1 \times \dots \times \mu_k$ 

where each  $\mu_i$  is a  $\sigma$ -finite measure on  $\mathcal{B}_i$ . Thus,  $\gamma$  is the probability distribution of  $(X_1, \ldots, X_k)$  and  $\mu$  a  $\sigma$ -finite product measure.

By Theorem 4, it follows that  $P|\mathcal{D}_0$  is atomic whenever  $\gamma \ll \mu$ . Whether or not  $\gamma \ll \mu$  implies  $P|\mathcal{D}$  atomic is a bit more delicate and is the main focus of this section. We start by noting that, in the independent case, things are as expected.

**Lemma 7.** Let  $A_i = \sigma(X_i^*)$  for all i and

$$\mathcal{D}_j = \bigcap_{i=1}^j \overline{\sigma(X_1, \dots, X_{i-1}, X_{i+1}, \dots, X_j)}, \quad j = 2, \dots, k.$$

If  $X_j$  is independent of  $(X_1, \ldots, X_{j-1})$ , then  $\mathcal{D}_j = \mathcal{D}_{j-1}$ . In particular, if  $X_1, \ldots, X_k$  are independent, then  $\mathcal{D} = \mathcal{N}$  and  $H = \{X_1 \in B_1, \ldots, X_k \in B_k\}$  is TIP as far as  $B_i \in \mathcal{B}_i$  for all i and P(H) > 0.

*Proof.* Since  $\mathcal{D}_{j-1} \subset \mathcal{D}_j$ , it suffices to prove  $\mathcal{D}_{j-1} \supset \mathcal{D}_j$ . Let  $A \in \mathcal{D}_j$ . Then,  $I_A = f_i(X_1, \dots, X_{i-1}, X_{i+1}, \dots, X_j)$  a.s. for some bounded measurable function  $f_i$ ,  $i = 1, \dots, j$ . Let  $\alpha_j$  denote the probability distribution of  $X_j$ . If  $X_j$  is independent of  $(X_1, \dots, X_{j-1})$ , then

$$I_A = E(I_A \mid X_1, \dots, X_{j-1})$$

$$= E(f_i(X_1, \dots, X_{i-1}, X_{i+1}, \dots, X_j) \mid X_1, \dots, X_{j-1})$$

$$= \int f_i(X_1, \dots, X_{i-1}, X_{i+1}, \dots, X_{j-1}, t) \alpha_j(dt) \text{ a.s. for each } i < j.$$

Thus,  $A \in \mathcal{D}_{j-1}$ . Next, suppose  $X_1, \ldots, X_k$  are independent. By what already proved,

$$\mathcal{D} = \mathcal{D}_k = \mathcal{D}_{k-1} = \ldots = \mathcal{D}_2 = \overline{\sigma(X_1)} \cap \overline{\sigma(X_2)} = \mathcal{N},$$

or equivalently  $\Omega$  is TIP. Since  $X_1, \ldots, X_k$  are still independent under  $P(\cdot \mid H)$ , it follows that  $\Omega$  is  $P(\cdot \mid H)$ -TIP, that is, H is P-TIP.

The independence assumption can be considerably relaxed. Next result is inspired to Corollary 3.7 of [3].

**Theorem 8.** Suppose  $A_i = \sigma(X_i^*)$  for all  $i, \gamma \ll \mu$  and f is a version of  $\frac{d\gamma}{d\mu}$ . Let  $H = \{(X_1, \ldots, X_k) \in B\}$  where  $B \in \mathcal{B}$  and  $B \subset \{f > 0\}$ .

Then, H is TIP provided

(5) 
$$\bigcup_{i=1}^{k} \{X_i^* \in B_i^*\} \supset H \supset \bigcap_{i=1}^{k} \{X_i \in B_i\}, \text{ where } B_i^* = \times_{j \neq i} B_j,$$
 for some  $B_i \in \mathcal{B}_i, i = 1, ..., k, \text{ with } P(X_1 \in B_1, ..., X_k \in B_k) > 0.$ 

Proof. It can be assumed  $(\Omega, \mathcal{F}, P) = (\mathcal{X}, \mathcal{B}, \gamma)$  and  $X_1, \ldots, X_k$  the canonical projections (so that H = B). For each i, since  $\mu_i$  is  $\sigma$ -finite, there is a probability  $Q_i$  on  $\mathcal{B}_i$  equivalent to  $\mu_i$ . Let  $Q = Q_1 \times \ldots \times Q_k$  denote the corresponding product probability on  $\mathcal{B}$ . Since  $P \ll Q$  and P(H) > 0, then Q(H) > 0. Since f > 0 on H, then  $Q(\cdot \mid H)$  is equivalent to  $P(\cdot \mid H)$ . Thus, H is P-TIP if and only if it is Q-TIP. We next prove that H is Q-TIP. Let  $K = \{X_1 \in B_1, \ldots, X_k \in B_k\}$ . Since Q(K) > 0 (due to P(K) > 0) and  $X_1, \ldots, X_k$  are independent under Q, Lemma 7 implies that K is Q-TIP. Fix  $A_i \in \mathcal{A}_i$  with  $Q(A_i \Delta A_1 \mid H) = 0$ ,  $i = 1, \ldots, k$ . Since  $Q(A_i \mid K) = 0$  for all  $Q(A_i \mid K) = 0$ 

 $Q(A_j, X_j^* \in B_j^*) Q(X_j \in B_j) = Q(A_j, X_j^* \in B_j^*, X_j \in B_j) = Q(A_j \cap K) = 0,$ then  $Q(A_j, X_j^* \in B_j^*) = 0$ . Also,  $\{X_j^* \in B_j^*\} \cap \{X_r \notin B_r\} = \emptyset$  for  $j \neq r$  and  $H \subset \bigcup_{i=1}^k \{X_i^* \in B_i^*\}$ . Thus, letting  $A = \bigcap_i A_i$ ,

$$Q(A_1 \cap H) = Q(A \cap H) = Q(A \cap H \cap K^c)$$

$$\leq Q(A \cap (\cup_j \{X_j^* \in B_j^*\}) \cap (\cup_r \{X_r \notin B_r\}))$$

$$\leq \sum_j \sum_r Q(A, X_j^* \in B_j^*, X_r \notin B_r)$$

$$= \sum_j Q(A, X_j^* \in B_j^*, X_j \notin B_j)$$

$$\leq \sum_j Q(A_j, X_j^* \in B_j^*) = 0.$$

Thus, H is Q-TIP, and this concludes the proof.

By Theorem 8,  $\mathcal{D} = \mathcal{N}$  in case  $\gamma \ll \mu$  and condition (5) holds with  $B = \{f > 0\}$ .

**Example 9.** Let  $X_3 = X_1 X_2$  where  $X_1$  and  $X_2$  are i.i.d. random variables with values in  $\{-1,1\}$  and  $P(X_1 = -1) = P(X_1 = 1) = \frac{1}{2}$ . Let  $\mu_1 = \mu_2 = \mu_3$  be counting measure on  $\{-1,1\}$  and  $\mathcal{D}_0 = \bigcap_i \overline{\sigma(X_i)}$ . Since the  $X_i$  are pairwise independent (even if not independent),  $\mathcal{D}_0 = \mathcal{N}$ . Since  $P(X_3 = 1) = \frac{1}{2}$  and  $\mathcal{D} \supset \sigma(X_3)$ , then  $\mathcal{D} \neq \mathcal{N}$ . Thus,  $\mathcal{D}_0 = \mathcal{N}$  and  $\gamma \ll \mu$  do not imply  $\mathcal{D} = \mathcal{N}$ . Note also that  $\bigcup_{i=1}^3 \{X_i = 1\} = \Omega$ ,  $P(X_1 = X_2 = X_3 = 1) > 0$  while  $H = \Omega$  is not TIP. Thus, condition (5) cannot be weakened into

$$\bigcup_{i=1}^{k} \{X_i \in B_i\} \supset H \supset \bigcap_{i=1}^{k} \{X_i \in B_i\} \text{ with } P(X_1 \in B_1, \dots, X_k \in B_k) > 0.$$

Our last and main result is that  $\mathcal{D}$  is atomic under P as far as  $\gamma \ll \mu$ .

**Theorem 10.** Let  $A_i = \sigma(X_i^*)$  for all i. If  $\gamma \ll \mu$ , then  $P|\mathcal{D}$  is atomic.

*Proof.* Let Q be a probability measure on  $\mathcal{F}$  which makes  $X_1, \ldots, X_k$  independent. Denote  $M_Q$  the class of those probabilities  $\mathbb{P}$  on  $\mathcal{F}$  such that  $\mathbb{P} \ll Q$  and

$$\mathcal{N}_{\mathbb{P}} = \{ F \in \mathcal{F} : \mathbb{P}(F) \in \{0,1\} \}, \quad \mathcal{D}_{\mathbb{P}} = \bigcap_{i=1}^{k} \sigma(\sigma(X_i^*) \cup \mathcal{N}_{\mathbb{P}}), \text{ with } \mathbb{P} \in M_Q.$$

Arguing by induction on k, we now prove that each  $\mathbb{P} \in M_Q$  is atomic on  $\mathcal{D}_{\mathbb{P}}$ .

Let k=2 and  $\mathbb{P} \in M_Q$ . Since  $X_1^*=X_2$  and  $X_2^*=X_1$ , then  $\mathbb{P}$  is atomic on  $\mathcal{D}_{\mathbb{P}}$  by Theorem 4.

Given  $k \geq 3$ , define  $V_i = (X_1, \dots, X_{i-1}, X_{i+1}, \dots, X_{k-1})$ . By induction, suppose that each  $\mathbb{P} \in M_Q$  is atomic on

$$\mathcal{V}_{\mathbb{P}} = \bigcap_{i=1}^{k-1} \sigma(\sigma(V_i) \cup \mathcal{N}_{\mathbb{P}}).$$

We have to prove that each  $\mathbb{P} \in M_Q$  is atomic on  $\mathcal{D}_{\mathbb{P}}$ . Accordingly, we fix  $\mathbb{P} \in M_Q$  and a  $\mathcal{D}_{\mathbb{P}}$ -measurable function  $U: \Omega \to \mathbb{R}$ . Arguing as in the proof of Theorem 4, it suffices to show that  $\mathbb{P}(U \in C) = 1$  for some countable set  $C \subset \mathbb{R}$ .

Since  $\sigma(U) \subset \mathcal{D}_{\mathbb{P}}$  and  $\mathcal{A}_i = \sigma(X_i^*)$ , one obtains  $\mathbb{P}(U = f_i(X_i^*)) = 1$ ,  $i = 1, \ldots, k$ , for some real measurable function  $f_i$  on  $(\prod_{j \neq i} \mathcal{X}_j, \prod_{j \neq i} \mathcal{B}_j)$ . Let

$$A_x = \{ f_1(V_1, x) = \dots = f_{k-1}(V_{k-1}, x) \}$$
 for  $x \in \mathcal{X}_k$ ,  
 $F(t, x) = Q(A_x \cap \{ f_1(V_1, x) \le t \})$  for  $t \in \mathbb{R}$  and  $x \in \mathcal{X}_k$ .

Since  $F(t,\cdot)$  is  $\mathcal{B}_k$ -measurable for fixed t, F is a real cadlag process on the measurable space  $(\mathcal{X}_k, \mathcal{B}_k)$ . Let  $J = \{(t,x) : F(t,x) > F(t-,x)\}$  be the jump set of F. By a well known result (see e.g. [9], Proposition 2.26), J is contained in a countable union of graphs, that is,

$$J \subset \cup_n \{(g_n(x), x) : x \in \mathcal{X}_k\}$$

for suitable  $\mathcal{B}_k$ -measurable functions  $g_n: \mathcal{X}_k \to \mathbb{R}, n = 1, 2, \dots$ 

Fix  $x \in \mathcal{X}_k$  with  $Q(A_x) > 0$  and define  $Q_x(\cdot) = Q(\cdot \mid A_x)$ . Then,  $Q_x$  is atomic on  $\mathcal{V}_{Q_x}$  (since  $Q_x \in M_Q$ ) and  $f_1(V_1, x)$  is  $\mathcal{V}_{Q_x}$ -measurable (since  $Q_x(A_x) = 1$ ). Thus,  $F(\cdot, x)$  is a purely jump function, that is,  $Q(A_x \cap \{f_1(V_1, x) \notin J_x\}) = 0$  where  $J_x = \{t : F(t, x) > F(t-, x)\}$ . Integrating over x yields

$$Q(f_1(X_1^*) = \dots = f_{k-1}(X_{k-1}^*))$$
  
=  $Q(f_1(X_1^*) = \dots = f_{k-1}(X_{k-1}^*) = g_n(X_k)$  for some  $n$ ).

Let  $C = \{c \in \mathbb{R} : Q(f_k(X_k^*) = c) > 0\}$ . Since  $f_k(X_k^*)$  and  $g_n(X_k)$  are independent under Q, then  $Q(f_k(X_k^*) \notin C, f_k(X_k^*) = g_n(X_k)) = 0$  for all n. Hence,

$$Q(f_k(X_k^*) \notin C \text{ and } f_1(X_1^*) = \dots = f_k(X_k^*))$$
  
 
$$\leq Q(f_k(X_k^*) \notin C \text{ and } f_k(X_k^*) = g_n(X_k) \text{ for some } n) = 0.$$

Therefore,  $\mathbb{P} \ll Q$  and  $\mathbb{P}(U = f_i(X_i^*)) = 1$  for all i = 1, ..., k imply

$$\mathbb{P}(U \in C) = 1 - \mathbb{P}(f_k(X_k^*) \notin C \text{ and } f_1(X_1^*) = \ldots = f_k(X_k^*)) = 1.$$

Since C is countable,  $\mathbb{P}$  is atomic on  $\mathcal{D}_{\mathbb{P}}$ . This concludes the induction argument and proves that each  $\mathbb{P} \in M_Q$  is atomic on  $\mathcal{D}_{\mathbb{P}}$ .

Finally, to prove P atomic on  $\mathcal{D} = \mathcal{D}_P$ , it can be assumed  $(\Omega, \mathcal{F}, P) = (\mathcal{X}, \mathcal{B}, \gamma)$  and  $X_1, \ldots, X_k$  the canonical projections. Also, since  $\mu$  is a  $\sigma$ -finite product measure,  $\mu$  is equivalent to some probability Q on  $\mathcal{B} = \mathcal{F}$  which makes  $X_1, \ldots, X_k$  independent. Hence,  $\gamma \ll \mu$  implies  $P = \gamma \ll Q$ . This concludes the proof.

Note that Theorem 4 could be obtained as a corollary of previous Theorem 10. However, Theorem 4 has been stated as an autonomous result, since it is a useful preliminary step toward Theorem 10.

Finally, the scope of Theorems 8 and 10 can be enlarged via Lemma 6. Following this route, sometimes, the assumption  $\gamma \ll \mu$  can be circumvented. Let  $Z_i: \mathcal{X} \to \mathcal{X}_i$  denote the *i*-th canonical projection and  $Z_i^* = (Z_1, \ldots, Z_{i-1}, Z_{i+1}, \ldots, Z_k)$ . Moreover, suppose  $\gamma$  is equivalent on rectangles to some probability  $\nu$  on  $\mathcal{B}$ , i.e.,  $\gamma(A) = 0 \Leftrightarrow \nu(A) = 0$  for each set A of the form  $A = \{Z_1^* \in C_1, \ldots, Z_k^* \in C_k\}$  with  $C_i \in \prod_{j \neq i} \mathcal{B}_j$  for all i. Then,  $\mathcal{D}$  is atomic under P provided  $\nu \ll \mu$ ; cfr. Lemma 6 and Theorem 10. Or else,  $\mathcal{D} = \mathcal{N}$  whenever  $\nu \ll \mu$  and

$$\bigcup_{i=1}^{k} \{Z_i^* \in B_i^*\} \supset \{\frac{d\nu}{d\mu} > 0\} \supset \bigcap_{i=1}^{k} \{Z_i \in B_i\}$$

for some  $B_1, \ldots, B_k$  such that  $\nu(Z_1 \in B_1, \ldots, Z_k \in B_k) > 0$ ; cfr. Lemma 6 and Theorem 8. Note also that, for k = 2, equivalence on rectangles reduces to

$$\gamma(B_1 \times B_2) = 0 \Leftrightarrow \nu(B_1 \times B_2) = 0$$
 whenever  $B_1 \in \mathcal{B}_1, B_2 \in \mathcal{B}_2$ .

As an example (suggested by an anonymous referee) suppose  $(X_n : n \ge 1)$  is an exchangeable sequence of real random variables with Ferguson-Dirichlet mixing measure. For k = 2,  $(X_1, X_2)$  is distributed as

$$\gamma(B_1 \times B_2) = a \,\beta(B_1) \,\beta(B_2) + (1 - a) \,\beta(B_1 \cap B_2)$$

where 0 < a < 1 and  $\beta$  is a probability on the real Borel sets. Then  $\mathcal{D} = \mathcal{N}$ , as  $\gamma$  is equivalent on rectangles to  $\beta \times \beta$ . However, if  $\beta$  is nonatomic,  $\gamma$  fails to be absolutely continuous with respect to any  $\sigma$ -finite product measure. It can be shown that, for every  $k \geq 2$ , one obtains  $\mathcal{D} = \mathcal{N}$  for  $(X_1, \ldots, X_k)$  as well.

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#### References

- [1] Berti P., Pratelli L., Rigo P. (2006) Asymptotic behaviour of the empirical process for exchangeable data, *Stoch. Proc. Appl.*, 116, 337-344.
- [2] Berti P., Pratelli L., Rigo P. (2007) Skorohod representation on a given probability space, *Prob. Theo. Rel. Fields*, 137, 277-288.
- [3] Berti P., Pratelli L., Rigo P. (2008) Trivial intersection of  $\sigma$ -fields and Gibbs sampling, Ann. Probab., 36, 2215-2234.
- [4] Bhaskara Rao K.P.S., Bhaskara Rao M. (1983) Theory of charges, Academic Press.
- [5] Burkholder D.L., Chow Y.S. (1961) Iterates of conditional expectation operators, *Proc. Amer. Math. Soc.*, 12, 490-495.
- [6] Delyon B., Delyon F. (1999) Generalization of von Neumann's spectral sets and integral representation of operators, *Bull. Soc. Math. Fr.*, 127, 25-41.
- [7] Diaconis P., Khare K., Saloff-Coste L. (2007) Stochastic alternating projections, *Preprint*, Dept. of Statistics, Stanford University, currently available at: http://www-stat.stanford.edu/~cgates/PERSI/papers/altproject-2.pdf
- [8] Halmos P.R. (1947) On the set of values of a finite measure, Bull. Amer. Math. Soc., 53, 138-141.
- [9] Karatzas I., Shreve S.E. (1991) Brownian motion and stochastic calculus (second edition), Springer.
- [10] San Martin E., Mouchart M., Rolin J.M. (2005) Ignorable common information, null sets and Basu's first theorem, Sankhya, 67, 674-698.

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