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**Atomic Intersection of σ -Fields
and Some of Its Consequences**

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ATOMIC INTERSECTION OF σ -FIELDS AND SOME OF ITS CONSEQUENCES

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ABSTRACT. Let (Ω, \mathcal{F}, P) be a probability space. For each $\mathcal{G} \subset \mathcal{F}$, define $\overline{\mathcal{G}}$ as the σ -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 $\mathcal{A}_1, \dots, \mathcal{A}_k \subset \mathcal{F}$ sub- σ -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, (Ω, \mathcal{F}, P) is a probability space, $\mathcal{A}_1, \dots, \mathcal{A}_k \subset \mathcal{F}$ sub- σ -fields, $k \geq 2$, and we let

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

As discussed in Section 3, the sub- σ -field

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

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 $\cap_{i=1}^k \overline{\mathcal{A}_i} = \cap_{i=1}^k \mathcal{A}_i$; 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, \dots, X_k are any random variables on (Ω, \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, \dots, X_k) is absolutely continuous with respect to a σ -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 $A_i \in \mathcal{A}_i$, $i = 1, \dots, k$. 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 $Q(\cdot | H)$ 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 $Q(\cdot | K)$ 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 $P(\cdot | D)$ 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- σ -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 Ω coincides a.s. with some \mathcal{G} -measurable function. Thus, if $U : \Omega \rightarrow \mathbb{R}$ is \mathcal{D} -measurable, then $U = U_i$ a.s. for some \mathcal{A}_i -measurable function $U_i : \Omega \rightarrow \mathbb{R}$, $i = 1, \dots, k$.

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 (Ω, \mathcal{F}, P) with values in the measurable space $(\mathcal{X}_i, \mathcal{B}_i)$, $i = 1, \dots, k$, 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 (Ω, \mathcal{F}, P) with values in the measurable space $(\mathcal{X}, \mathcal{E})$. The intersection property (IP) is

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

where the notation " $U \perp V | W$ " 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 $\mathcal{A}_i = \sigma(X_i)$ for all i . Then:*

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

$$P(X \in A, X_1 \in B_1) = P(X \in A)P(X_1 \in B_1) \quad \text{whenever}$$

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

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

$$E(f(X) \mid X_i) = E(f(X) \mid \mathcal{D}) = E(f(X) \mid X_1, \dots, X_k) \text{ a.s.}$$

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 ,

$$E(f(X) \mid X_1) = E(f(X) \mid X_1, \dots, X_k) = E(f(X) \mid X_i) \quad \text{a.s. for all } i.$$

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) " \Leftarrow " is obvious. As to " \Rightarrow ", it suffices noting that $E(f(X) \mid X_1, \dots, X_k) = E(f(X) \mid \mathcal{D}) = E f(X)$ 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 $X \perp (X_1, \dots, X_k)$. Heuristically, $X \perp \mathcal{D}$ means that X is not affected by that part of information which is common to X_1, \dots, X_k .

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

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

As shown in Theorem 4, for $P|_{\mathcal{D}}$ to be atomic, it is enough that the distribution of (X_1, \dots, X_k) is absolutely continuous with respect to a σ -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 (Ω, \mathcal{F}, P) such that $EX^2 < \infty$ and

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

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

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

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, $E(X | \mathcal{D}) = \sum_j I_{D_j} E(X | D_j)$ a.s. where D_1, D_2, \dots denote the (disjoint) atoms of $P|_{\mathcal{D}}$. Thus, one should apply relation (1) on each atom D_j , so as to obtain an estimate for $E(X | D_j)$, and then use the formula $EX = \sum_j P(D_j)E(X | D_j)$.

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, \dots, X_k be the canonical projections on

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

Each X_i is assumed to admit a regular conditional distribution γ_i given X_i^* . In the notation $u = (x_1, \dots, x_{i-1}, x_{i+1}, \dots, x_k)$, 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) $P(B) = \int \int I_B(x_1, \dots, x_k) \gamma_i(u) dx_i \gamma_i^*(du)$ for $B \in \prod_{i=1}^k \mathcal{B}_i$, where γ_i^* denotes the marginal distribution of X_i^* .

The Gibbs-chain

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

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

$$\begin{aligned} K(\omega, B) &= K((x_1, \dots, x_k), B) \\ &= \int \dots \int I_B(a_1, \dots, a_k) \prod_{i=1}^k \gamma_i(x_1, \dots, x_{i-1}, a_{i+1}, \dots, a_k) (da_i). \end{aligned}$$

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

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

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

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

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

Now, letting $\mathcal{A}_i = \sigma(X_i^*)$ 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 σ -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, \dots), then (2) turns into

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

\mathbb{P} -a.s. on $\{Y_0 \in D_j\}$, for all j 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 $\int \phi(\omega) P(d\omega | D_j)$, and then the formula $\int \phi dP = \sum_j P(D_j) \int \phi(\omega) P(d\omega | D_j)$ should be applied.

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

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

Here are some obvious consequences of the definition.

(i) If H is TIP, $A_i \in \mathcal{A}_i$ and $P(A_i \Delta A_1 | H) = 0$ for all i , then either $P(A_i | H) = 0$ for all i or $P(A_i | H) = 1$ for all i .

(ii) Let $H \in \mathcal{F}$ with $P(H) > 0$ and write $P_H = P(\cdot | H)$. Then, H is TIP if and only if Ω is P_H -TIP (i.e., Ω is TIP under P_H). Moreover, Ω 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 $\mathcal{D}_{P_H} = \bigcap_{i=1}^k \sigma(\mathcal{A}_i \cup \mathcal{N}_{P_H})$.

(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 $P(D | H^*) = P(D | H)$ 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 $P(\bigcup_n H_n) = 1$ for some countable collection H_1, H_2, \dots of TIP sets.*

Proof. (a) Suppose H is TIP. We first prove that $P(\cdot | H)$ 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 $P(D | H) = P(A_1 | H) \in \{0, 1\}$. 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 $P(D | H) = 1$, 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, $P(D | H^*) = 1$. Taking complements, if $P(D | H) = 0$ then $P(D | H^*) = 0$. Thus, H^* is an atom of $P|\mathcal{D}$ and $P(\cdot | H^*) = P(\cdot | H)$ on \mathcal{D} .

(b) If $H \in \mathcal{D}$ is TIP, then H is an atom of $P|\mathcal{D}$ since $P(H | H^*) = P(H | H) = 1$, 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 $P(A_i \Delta A_1 | H) = 0$ for $i = 1, \dots, k$. 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,

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

Hence, $A_1 \cap H \in \overline{\mathcal{A}_i}$ for all i , that is, $A_1 \cap H \in \mathcal{D}$. Since H is an atom of $P|\mathcal{D}$, it follows that $P(A_1 | H) = P(A_1 \cap H | H) \in \{0, 1\}$. 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 $P(\cup_n H_n^*) \geq P(\cup_n H_n) = 1$. \square

By Theorem 2, $P|\mathcal{D}$ is atomic provided Ω can be covered by countably many TIP sets H_1, H_2, \dots . In this case, every atom D admits the representation $D = \cup_{i \in I} H_i$ 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, \dots are TIP, and let $D \subset \Omega$. Then, D is an atom of $P|\mathcal{D}$ if and only if D is TIP and*

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

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 $A_{i,n} \in \mathcal{A}_i$, $i = 1, \dots, k$, satisfying

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

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

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

Define $F_{i,n} = A_{i,n}$ or $F_{i,n} = A_{i,n}^c$ as $P(A_{i,n} | D) = 1$ or $P(A_{i,n} | D) = 0$, and

$$A_i = \cap_{n \in N} F_{i,n}.$$

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

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

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 $P(H_n \setminus D) = 0$. \square

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

Theorem 4. *$P \mid \mathcal{D}$ 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 $U : \Omega \rightarrow \mathbb{R}$ 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 $U : \Omega \rightarrow \mathbb{R}$ be \mathcal{D} -measurable. Then, $U = U_i$ a.s. for some $U_i : \Omega \rightarrow \mathbb{R}$ satisfying $\sigma(U_i) \subset \mathcal{A}_i$, $i = 1, \dots, k$. 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

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

\square

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

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

$$(4) \quad P(X_i = f(X_j)) = 0$$

for all $i \neq j$ and all measurable functions $f : \mathcal{X}_j \rightarrow \mathcal{X}_i$. We mention this fact since, for some time, we conjectured $P \mid \mathcal{D}$ 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 $f = (f_1, f_2) : \mathbb{R}^2 \rightarrow \mathbb{R}^2$, 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, $P \mid \mathcal{D}$ 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

$$P(A) = 0 \Leftrightarrow Q(A) = 0 \text{ for each } A \in \mathcal{R},$$

$$\text{where } \mathcal{R} = \{\cap_{i=1}^k A_i : A_i \in \mathcal{A}_i, i = 1, \dots, k\}.$$

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, Q are equivalent on rectangles. Note that $\mathcal{A}_i \subset \mathcal{R}$ 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,*

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

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

(Here, $\mathcal{N}_Q = \{F \in \mathcal{F} : Q(F) \in \{0, 1\}\}$, $\mathcal{D}_Q = \cap_{i=1}^k \sigma(\mathcal{A}_i \cup \mathcal{N}_Q)$, $\mathcal{N}_P = \mathcal{N}$ 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 $Q(A_i\Delta D) = 0$, $i = 1, \dots, k$, 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 $P(A \cap G^c) = P(A \cap G_1^c) = 0$, i.e., A is an atom of $P|\mathcal{D}_P$. Next, if $\mathcal{D}_Q = \mathcal{N}_Q$, then Ω 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, \dots . 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 $A \in \mathcal{A}_1 \subset \mathcal{R}$. Therefore, $P|\mathcal{D}_P$ is atomic. \square

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 (Ω, \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 σ -field on \mathcal{X} . Define two measures on \mathcal{B} as

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

where each μ_i is a σ -finite measure on \mathcal{B}_i . Thus, γ is the probability distribution of (X_1, \dots, X_k) and μ a σ -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 $\mathcal{A}_i = \sigma(X_i^*)$ for all i and*

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

If X_j is independent of (X_1, \dots, X_{j-1}) , then $\mathcal{D}_j = \mathcal{D}_{j-1}$. In particular, if X_1, \dots, X_k are independent, then $\mathcal{D} = \mathcal{N}$ and $H = \{X_1 \in B_1, \dots, X_k \in B_k\}$ 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 α_j denote the probability distribution of X_j . If X_j is independent of (X_1, \dots, X_{j-1}) , then

$$\begin{aligned} I_A &= E(I_A \mid X_1, \dots, X_{j-1}) \\ &= E(f_i(X_1, \dots, X_{i-1}, X_{i+1}, \dots, X_j) \mid X_1, \dots, X_{j-1}) \\ &= \int f_i(X_1, \dots, X_{i-1}, X_{i+1}, \dots, X_{j-1}, t) \alpha_j(dt) \quad \text{a.s. for each } i < j. \end{aligned}$$

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

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

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

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

Theorem 8. *Suppose $\mathcal{A}_i = \sigma(X_i^*)$ for all i , $\gamma \ll \mu$ and f is a version of $\frac{d\gamma}{d\mu}$. Let*

$$H = \{(X_1, \dots, X_k) \in B\} \quad \text{where } B \in \mathcal{B} \text{ and } B \subset \{f > 0\}.$$

Then, H is TIP provided

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

Proof. It can be assumed $(\Omega, \mathcal{F}, P) = (\mathcal{X}, \mathcal{B}, \gamma)$ and X_1, \dots, X_k the canonical projections (so that $H = B$). For each i , since μ_i is σ -finite, there is a probability Q_i on \mathcal{B}_i equivalent to μ_i . Let $Q = Q_1 \times \dots \times Q_k$ 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 $K = \{X_1 \in B_1, \dots, X_k \in B_k\}$. Since $Q(K) > 0$ (due to $P(K) > 0$) and X_1, \dots, X_k are independent under Q , Lemma 7 implies that K is Q -TIP. Fix $A_i \in \mathcal{A}_i$ with $Q(A_i \Delta A_1 \mid H) = 0$, $i = 1, \dots, k$. Since K is Q -TIP and $K \subset H$, then $Q(A_1 \mid K) \in \{0, 1\}$, say $Q(A_1 \mid K) = 0$ (so that $Q(A_i \mid K) = 0$ for all i). Given j , since $Q(X_j \in B_j) > 0$ and

$$Q(A_j, X_j^* \in B_j^*) Q(X_j \in B_j) = Q(A_j, X_j^* \in B_j^*, X_j \in B_j) = Q(A_j \cap K) = 0,$$

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 $H \subset \cup_{i=1}^k \{X_i^* \in B_i^*\}$. Thus, letting $A = \cap_i A_i$,

$$\begin{aligned} Q(A_1 \cap H) &= Q(A \cap H) = Q(A \cap H \cap K^c) \\ &\leq Q(A \cap (\cup_j \{X_j^* \in B_j^*\}) \cap (\cup_r \{X_r \notin B_r\})) \\ &\leq \sum_j \sum_r Q(A, X_j^* \in B_j^*, X_r \notin B_r) \\ &= \sum_j Q(A, X_j^* \in B_j^*, X_j \notin B_j) \\ &\leq \sum_j Q(A_j, X_j^* \in B_j^*) = 0. \end{aligned}$$

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

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 $\mathcal{D}_0 = \cap_i \overline{\sigma(X_i)}$. 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 $\cup_{i=1}^3 \{X_i = 1\} = \Omega$, $P(X_1 = X_2 = X_3 = 1) > 0$ while $H = \Omega$ is not TIP. Thus, condition (5) cannot be weakened into

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

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

Theorem 10. *Let $\mathcal{A}_i = \sigma(X_i^*)$ 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, \dots, X_k independent. Denote M_Q the class of those probabilities \mathbb{P} on \mathcal{F} such that $\mathbb{P} \ll Q$ and

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

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

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

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 $U : \Omega \rightarrow \mathbb{R}$. 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 $\mathbb{P}(U = f_i(X_i^*)) = 1$, $i = 1, \dots, k$, for some real measurable function f_i on $(\prod_{j \neq i} \mathcal{X}_j, \prod_{j \neq i} \mathcal{B}_j)$. Let

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

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 $g_n : \mathcal{X}_k \rightarrow \mathbb{R}$, $n = 1, 2, \dots$

Fix $x \in \mathcal{X}_k$ with $Q(A_x) > 0$ and define $Q_x(\cdot) = Q(\cdot | A_x)$. 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

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

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,

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

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

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

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, \dots, X_k the canonical projections. Also, since μ is a σ -finite product measure, μ is equivalent to some probability Q on $\mathcal{B} = \mathcal{F}$ which makes X_1, \dots, X_k independent. Hence, $\gamma \ll \mu$ implies $P = \gamma \ll Q$. This concludes the proof. \square

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 $Z_i : \mathcal{X} \rightarrow \mathcal{X}_i$ denote the i -th canonical projection and $Z_i^* = (Z_1, \dots, Z_{i-1}, Z_{i+1}, \dots, Z_k)$. Moreover, suppose γ is equivalent on rectangles to some probability ν on \mathcal{B} , i.e., $\gamma(A) = 0 \Leftrightarrow \nu(A) = 0$ for each set A of the form $A = \{Z_1^* \in C_1, \dots, Z_k^* \in C_k\}$ 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

$$\cup_{i=1}^k \{Z_i^* \in B_i^*\} \supset \left\{ \frac{d\nu}{d\mu} > 0 \right\} \supset \cap_{i=1}^k \{Z_i \in B_i\}$$

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

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

As an example (suggested by an anonymous referee) suppose $(X_n : n \geq 1)$ 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 β is a probability on the real Borel sets. Then $\mathcal{D} = \mathcal{N}$, as γ is equivalent on rectangles to $\beta \times \beta$. However, if β is nonatomic, γ fails to be absolutely continuous with respect to any σ -finite product measure. It can be shown that, for every $k \geq 2$, one obtains $\mathcal{D} = \mathcal{N}$ for (X_1, \dots, 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 σ -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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