

CONDITIONAL PROBABILITY IN NONSTANDARD ANALYSIS

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Abstract. In this paper we apply the theory of Loeb measure to conditional probability for hyperfinite Loeb spaces. We show that conditional probability $\sim P(\cdot/A)$ on a Loeb space $(V, \mathfrak{M}(\sim P), \sim P)$ for $A \in {}^* \mathfrak{P}(V)$ ($P(A) > 0$ and $P(A) \approx' 0^1$ is a Loeb measure and for $A \in \mathfrak{M}(\sim P)$ ($\sim P(A) > 0$) can be represented by a Loeb measure. For the case $A \in \mathfrak{M}(\sim P)$ we prove that there exists a set $C \in {}^* \mathfrak{P}(V)$ such that $\sim P(\cdot/A)$ is equal to the Loeb conditional probability $L(P(\cdot/C))$. We introduce internal conditional probability relative to an internal subalgebra \mathfrak{A} of ${}^* \mathfrak{P}(V)$ as in case of finite standard probability spaces. We show, analogously to a well-known probability result, that internal conditional probability $P(A/\mathfrak{A})$, $A \in {}^* \mathfrak{P}(V)$, and internal conditional expectation $E(X/\mathfrak{A})$, X is S -integrable, are P -a. s. unique, in nonstandard sense, random variables on (V, \mathfrak{A}, P) . Finally, we give a nonstandard characterization of conditional probability $\sim P(A/\mathfrak{M}(\mathfrak{A}))$, $A \in \mathfrak{M}(\sim P)$ on a Loeb space $(V, \mathfrak{M}(\sim P), \sim P)$. We prove that there exists a set $C \in {}^* \mathfrak{P}(V)$ such that $P(C/\mathfrak{A})$ is the lifting of $\sim P(A/\mathfrak{M}(\mathfrak{A}))$.

Introduction. In this paper we concern ourselves with conditional probability for hyperfinite Loeb spaces. We use the well-known results from the theory of Loeb measure [8] and nonstandard probability [2], [10] and the methodology developed by P. Loeb, J. Keisler, R. Anderson and others.

In the first part we define internal conditional probability $P(\cdot/A)$, $A \in {}^* \mathfrak{P}(V)$ for a hyperfinite probability space $(V, {}^* \mathfrak{P}(V), P)$ and give the nonstandard representation of conditional probability $\sim P(\cdot/A)$, $A \in \mathfrak{M}(\sim P)$ on the Loeb space $(V, \mathfrak{M}(\sim P), \sim P)$. We show that for $A \in {}^* \mathfrak{P}(V)$ with $P(A) > 0$ and $P(A) \approx' 0$ $\sim P(\cdot/A)$ is a Loeb measure on $(V, \mathfrak{M}(\sim P))$ and for $A \in \mathfrak{M}(\sim P)$ with $\sim P(A) > 0$ there exists a set $C \in {}^* \mathfrak{P}(V)$ such that $\sim P(\cdot/A)$ can be represented by the Loeb conditional probability $L(P(\cdot/C))$.

In the second part we define, analogously to the definition of internal conditional expectation $E(X/\mathfrak{A})$, [10], internal conditional probability $P(A/\mathfrak{A})$ $A \in {}^* \mathfrak{P}(V)$, is an internal subalgebra of ${}^* \mathfrak{P}(V)$) for a hyperfinite probability space $(V, {}^* \mathfrak{P}(V), P)$. We show that so-introduced $P(A/\mathfrak{A})$ ($E(X/\mathfrak{A})$ as well) is

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¹ \approx' is the negation of \approx

P -a. s. unique random variable on (V, \mathfrak{A}, P) . The P -a. s. uniqueness of the $P(A/\mathfrak{A})$ ($E(X/\mathfrak{A})$) is introduced in theorem 4, corresponding to the same concept from standard probability. Finally, we give a nonstandard characterization of conditional probability $P(A/\mathfrak{M}(\mathfrak{A}))$ ($A \in \mathfrak{M}(\sim P)$, $(A \in \mathfrak{M}(\sim P) \mathfrak{M}(\mathfrak{A})$ is a sub- σ -algebra of $\mathfrak{M}(\sim P)$) on a Loeb space $(V, \mathfrak{M}(\sim P), \sim P)$. We show that for $A \in \mathfrak{M}(\sim P)$ there exists a set $B \in {}^* \mathfrak{P}(V)$ such that $P(B/\mathfrak{A})$ is lifting of $\sim P(A/\mathfrak{M}(\mathfrak{A}))$.

We assume $(V, {}^* \mathfrak{P}(V), P)$ to be a hyperfinite probability space and $(V, \mathfrak{M}(\sim P), \sim P)$ a Loeb space constructed from it, i.e. Loeb measure $\sim P$ is defined by

$$\begin{aligned} \sim P(F) &= \inf\{\text{st}(P(A)) \mid F \subseteq A \text{ and } A \in {}^* \mathfrak{P}(V)\} \\ &= \sup\{\text{st}(P(A)) \mid A \subseteq F \text{ and } A \in {}^* \mathfrak{P}(V)\} \end{aligned} \quad \text{for } F \subseteq V$$

and $\mathfrak{M}(\sim P)$ is a σ -algebra of all $\sim P$ -measurable sets $F \subseteq V$.

According to standard probability, for $A \in {}^* \mathfrak{P}(V)$ with $P(A) > 0$ we define internal conditional probability $P(\cdot/A)$ of an internal event B relative to A by

$$P(B/A) = P(A \cap B)/P(A)$$

It is easy to show that $(V, {}^* \mathfrak{P}(V), P(\cdot/A))$ is a hyperfinite probability space, so it gives rise to a Loeb space denoted by $(V, \mathfrak{M}(L(P(\cdot/A))), L(P(\cdot/A)))$.

On the other hand, for a Loeb space $(V, \mathfrak{M}(\sim P), \sim P)$, conditional probability $\sim P(\cdot/A)$ of an event $B \in \mathfrak{M}(\sim P)$ relative to $A \in \mathfrak{M}(\sim P)$, $\sim P(A) > 0$, is standardly defined by

$$\sim P(B/A) = \sim P(A \cap B)/\sim P(A)$$

It is well known that $\sim P(\cdot/A)$ is a probability measure on $(V, \mathfrak{M}(\sim P))$ but not necessary a Loeb measure. However, for $A \in {}^* \mathfrak{P}(V)$ with $P(A) > 0$ $P(A) \approx' 0$, $\sim P(\cdot/A)$ is a Loeb measure and we shall show it in this paper.

Let $\sigma(\sim P(\cdot/A))$ be σ -algebra of all $\sim P(\cdot/A)$ -measurable sets, i.e.

$$\sigma(\sim P(\cdot/A)) = \{F \subseteq V \mid F \cap A \in \mathfrak{M}(\sim P)\}$$

It is obvious that $\mathfrak{M}(\sim P) \subset \sigma(\sim P(\cdot/A))$. From the theory of Loeb measure we know that $\sim P$ is a complete measure on $(V, \mathfrak{M}(\sim P))$ and that $\mathfrak{M}(\sim P)$ is a completion of $\mathcal{L}(V)$ relative to $\sim P$. We have the same for $\sim P(\cdot/A)$:

LEMMA 1. *Probability measure $\sim P(\cdot/A)$ is a complete measure on $(V, \mathfrak{M}(\sim P))$ and $\mathfrak{M}(\sim P)$ is a completion of $\mathcal{L}(V)$ relative to $\sim P(\cdot/A)$ in the sense that for $F \in \mathfrak{M}(\sim P)$ there exist sets $Z \in \mathcal{L}(V)$ and $N \subseteq V$ such that*

$$F = Z \cup N, \quad N \subseteq U, \quad \text{and} \quad \sim P(U/A) = 0.$$

Proof Let $F \in \mathfrak{M}(\sim P)$, $\sim P(F/A) = 0$ and $M \subseteq F$. Then is $F \cap A \in \mathfrak{M}(\sim P)$, $M \cap A \subseteq F \cap A$ and $\sim P(F \cap A) = 0$. Since $\sim P$ is a complete measure, $M \cap A \in \mathfrak{M}(\sim P)$ and $\sim P(M \cap A) = 0$. This implies that M is $\sim P(\cdot/A)$ -measurable and $\sim P(M/A) = 0$.

Let $F \in \mathfrak{M}(\sim P)$. Then there exist sets $Z \in \mathbb{L}_\nu(V)$ and $N \subseteq V$ such that $F = Z \cup N$, $N \subseteq U$ and $\sim P(U) = 0$. Since $U \cap A \subseteq U$ and $\sim P$ is a complete measure, it follows that $\sim P(U \cap A) = 0$, i.e. $\sim P(U/A) = 0$. Hence $\mathfrak{M}(\sim P)$ is a completion of $\mathbb{L}_\nu(V)$ relative to $\sim P(\cdot/A)$.

The next theorem shows that conditional probability $\sim P(\cdot/A)$ for $A \in^* \mathfrak{P}(V)$, $P(A) > 0$, and $P(A) \approx' 0$ is a Loeb measure on $(V, \mathfrak{M}(\sim P))$.

THEOREM 1. *Let $A \in^* \mathfrak{P}(V)$, $P(A) > 0$ and $P(A) \approx' 0$. Then $(V, \mathfrak{M}(\sim P), \sim P(\cdot/A))$ is a Loeb probability space.*

Proof. We show that $\sim P(\cdot/A)$ is a Loeb measure obtained from internal conditional probability $P(\cdot/A)$. Using notations already defined we prove that

$$L(P(B/A)) = \sim P(B/A) \quad \text{for } B \in \mathfrak{M}(\sim P) \quad (1)$$

Let $F \in^* \mathfrak{P}(V)$. Since $P(A) > 0$ and $P(A) \approx' 0$

$$\begin{aligned} \sim P(F/A) &= \sim P(F \cap A)/\sim P(A) = \text{st}(P(F \cap A))/\text{st}(P(A)) = \\ &= \text{st}(P(F \cap A)/P(A)) = L(P(F/A)). \end{aligned}$$

Let $F \in \mathbb{L}_\nu(V)$. The definition of Loeb measure

$$\begin{aligned} L(P(F/A)) &= \\ &= \sup\{\text{st}(P(C/A)) \mid C \in^* \mathfrak{P}(V), C \subseteq F\} = \inf\{\text{st}(P(D/A)) \mid (D \in^* \mathfrak{P}(V), D \supseteq F)\} \end{aligned}$$

implies that for $\varepsilon \in^\sigma R^+$ there exist sets $C, D \in^* \mathfrak{P}(V)$ such that

$$\sim P(C/A) \leq \sim P(F/A) \leq \sim P(D/A) \quad \text{and} \quad (2)$$

$$\sim P(D/A) - \varepsilon < L(P(F/A)) < \sim P(C/A) + \varepsilon \quad (3)$$

Relations (2) and (3) imply

$$\sim P(F/A) - \varepsilon < L(P(F/A)) < \sim P(F/A) + \varepsilon \quad \text{i.e.} \quad L(P(F/A)) = \sim P(F/A)$$

Let $F \in \mathfrak{M}(\sim P)$. Then, according to [8], there exist sets $C, D \in \mathbb{L}_\nu(V)$ such that $C \subseteq F \subseteq D$ and $\sim P(C) = \sim P(F) = \sim P(D)$. We show that $F \in \mathfrak{M}(L(P(\cdot/A)))$ and that $L(P(F/A)) = \sim P(F/A)$. Since

$$\begin{aligned} \sim P(D \cup A) &= \sim P(D) + \sim P(A) - \sim P(D \cap A) \leq \sim P(D) + \sim P(A) - \sim P(C \cap A) = \\ &= \sim P(C) + \sim P(A) - \sim P(C \cap A) = \sim P(C \cup A) \leq \sim P(D \cup A) \end{aligned}$$

it follows that $\sim P(D \cap A) = \sim P(C \cap A)$, whence, and from $\sim P(C \cap A) \leq \sim P(F \cap A) \leq \sim P(D \cap A)$ we get

$$\sim P(C/A) = \sim P(F/A) = \sim P(D/A) \quad (4)$$

From (1) and (4) it follows that

$$L(P(C/A)) = L(P(D/A)) = \sim P(F/A) \quad (5)$$

and $L(P(DC/A)) = 0$. For the set $F \in \mathfrak{M}(\sim P)$ we have the following representation:

$$F = C \cup (FC) \text{ where } C \in \mathfrak{L}(V), FC \subseteq DC \text{ and } L(P(DC/A)) = 0.$$

So $F \in \mathfrak{M}(L(P(\cdot/A)))$. Since $C \subseteq F \subseteq D$ and $C, D, F \in \mathfrak{M}(L(P(\cdot/A)))$ we have that $L(P(C/A)) \leq L(P(F/A)) \leq L(P(D/A))$ which, in view of (5), implies

$$\sim P(F/A) \leq L(P(F/A)) \leq \sim P(F/A) \quad \text{i.e.} \quad L(P(F/A)) = \sim P(F/A)$$

Later on, whenever $A \in^* \mathfrak{P}(V)$, $P(A) > 0$ and $P(A) \approx' 0$, the conditional probability $\sim P(\cdot/A)$ on $(V, \mathfrak{M}(\sim P))$ will be denoted by $L(P(\cdot/A))$; assuming that it is a Loeb measure.

We now prove a representation theorem for conditional probability $\sim P(\cdot/A)$ ($A \in \mathfrak{M}(\sim P)$ and $\sim P(A) > 0$) on $(V, \mathfrak{M}(\sim P))$. We shall show that there exists a set $C \in^* \mathfrak{P}(V)$ with $P(C) > 0$ and $P(C) \approx' 0$ such that the conditional probability $\sim P(\cdot/A)$ is equal to the Loeb conditional probability $L(P(\cdot/C))$.

THEOREM 2. *Let $A \in \mathfrak{M}(\sim P)$ with $\sim P(A) > 0$. Then, there exists a set $C \in^* \mathfrak{P}(V)$ with $P(C) > 0$ and $P(C) \approx' 0$ such that*

$$L(P(F/C)) = \sim P(F/A), \quad \text{for any } F \in \mathfrak{M}(\sim P)$$

Proof. According to [8] exists a set $C \in^* \mathfrak{P}(V)$ such that $\sim P(C\Delta A) = 0$. We show that $P(C) > 0$ and $P(C) \approx' 0$: For sets $A, C \subseteq V$ we have that $C \setminus A \subseteq C\Delta A$, and $A \setminus C \subseteq C\Delta A$, so, by completeness of measure $\sim P$

$$\sim P(C \setminus A) = \sim P(A \setminus C) = 0$$

Since $C = (C \setminus A) \cup (C \cap A)$ and $A = (A \setminus C) \cup (A \cap C)$ and sets A, C satisfy (1)

$$\sim P(C) = \sim P(C \setminus A) + \sim P(C \cap A) = \sim P(C \cap A) = \sim P(A \setminus C) + (C \cap A) = \sim P(A) \quad (2)$$

Hence $P(C) > 0$ and $P(C) \approx' 0$,

Let $F \in \mathfrak{M}(\sim P)$. Then

$$F \cap C = (F \cap A \cap C) \cup ((C \setminus A) \cap F) \text{ and } F \cap A = (F \cap A \cap C) \cup ((A \setminus C) \cap F) \quad (3)$$

From (3), $(C \setminus A) \cap F \subseteq A\Delta C$, $(A \setminus C) \cap F \subseteq A\Delta C$, $\sim P((C \setminus A) \cap F) = 0$ and $\sim P((A \setminus C) \cap F) = 0$ it follows that

$$\begin{aligned} \sim P((A \setminus C) \cap F) &= \sim P(F \cap A \cap C) + \sim P((C \setminus A) \cap F) = \sim P(F \cap A \cap C) = \\ &= \sim P(F \cap A \cap C) + \sim P((A \setminus C) \cap F) = \sim P(F \cap A) \end{aligned} \quad (4)$$

Finally, according to theorem 1 (2) and (4) imply

$$L(P(F/C)) = \sim P(F/C) = \sim P(F \cap C) / \sim P(C) = \sim P(F \cap A) / \sim P(A) = \sim P(F/A)$$

The following theorem is a simple consequence of the Loeb theorem [8], but can be quite useful when working in nonstandard probability.

THEOREM 3. *Let $A \in \mathfrak{M}(\sim P)$ with $\sim P(A) > 0$. Then, for any set $F \in \mathfrak{M}(\sim P)$ there exists a set $C \in {}^* \mathfrak{P}(V)$ such that $\sim P(F/A) = \sim P(C/A)$.*

Proof. For $F \in \mathfrak{M}(\sim P)$, by the Loeb theorem [8], there exists a set $C \in {}^* \mathfrak{P}(V)$ such that $\sim P(F \Delta C) = 0$. Since $F \cap A = (F \cap A \cap C) \cup ((F \setminus C) \cap A)$ $C \cap A = (F \cap A \cap C) \cup ((C \setminus F) \cap A)$ $(F \setminus X) \cap A \subseteq F \Delta C$ and $(C \setminus F) \cap A \subseteq F \Delta C$, by the same arguments as in theorem 2, we get that $\sim P(F \cap A) = \sim P(C \cap A)$, i.e. $\sim P(F/A) = \sim P(C/A)$.

In the second part of this paper we are dealing with internal conditional expectation $E(X/\mathfrak{A})$ of an internal random variable $X : V \rightarrow {}^* R$ relative to \mathfrak{A} , where \mathfrak{A} is an internal subalgebra of ${}^* \mathfrak{P}(V)$ and $P(A/\mathfrak{A})$ denotes the internal conditional probability of an event $A \in {}^* \mathfrak{P}(V)$ relative to \mathfrak{A} .

We consider a hiperfinite probability space $(V, {}^* \mathfrak{P}(V), P)$, $A \in {}^* \mathfrak{P}(V)$ and internal subalgebra of ${}^* \mathfrak{P}(V)$. The hyperfiniteness of \mathfrak{A} implies, by transfer principle, that \mathfrak{A} is generated by a hyperfinite partition $\{V_1, V_2, \dots, V_H\}$ ($H \in {}^* N \setminus N$) of the set V . It permits us to define $P(A/\mathfrak{A})$ in the same way as in the case of finite standard probability spaces:

$$P(A/\mathfrak{A})(v) = \sum_{i=1}^H P(A/V_i) I_{V_i}(v) \quad \text{for } v \in V \quad (1)$$

where $P(A/V_i) = P(A \cap V_i)/P(V_i)$ $i = 1, 2, \dots, H$. Since

$$\begin{aligned} P(A/\mathfrak{A})(v) &= \sum_{i=1}^H P(A/V_i) I_{V_i}(v) \\ &= \sum_{i=1}^H (P(A \cap V_i)/P(V_i)) I_{V_i}(v) \\ &= \sum_{i=1}^H ((P(V_i))^{-1} \sum_{u \in V_i} (P(u) I_A(u))) I_{V_i}(v) \\ &= \sum_{i=1}^H E(I_A/V_i) I_{V_i}(v) \\ &= E(I_A/\mathfrak{A})(v) \quad \text{i.e.} \end{aligned}$$

$$P(A/\mathfrak{A}) = E(I_A/\mathfrak{A}). \quad (2)$$

in the further work we shall use both (1) and (2) as definitions of internal conditional probability.

For internal random variable $X : V \rightarrow {}^* R$ on $(V, {}^* \mathfrak{P}(V), P)$ internal conditional expectation $E(X/\mathfrak{A})$ has already been defined [6]. In [10] it is proved

that $E(X/\mathfrak{A})$ is an S - \mathfrak{A} -integrable random variable on (V, \mathfrak{A}, P) provided X is S -integrable. This result applied to $P(A/\mathfrak{A})$ implies that $P(A/\mathfrak{A})$ is an S - \mathfrak{A} -integrable random variable on (V, \mathfrak{A}, P) since I_A is S -integrable, [9]. In [10] it is proved that

$$E(E(X/\mathfrak{A})) = E(X). \quad (\text{i})$$

Taking (2) as definition of $P(A/\mathfrak{A})$, from (i) it follows, [9], that

$$E(P(A/\mathfrak{A})) = P(A) \quad (\text{ii})$$

Results (i) and (ii) make the Theorem of probability completeness for $E(X/\mathfrak{A})$ and $P(A/\mathfrak{A})$ hold for hyperfinite probability spaces.

We now prove a nonstandard version of the well known probability theorem, namely, that conditional probability and expectation relative to σ -subalgebra \mathfrak{B} are μ -a. s. unique random variables on (V, \mathfrak{B}, μ) , [11].

THEOREM 4. *Let $(V, {}^* \mathfrak{P}(V), P)$ be a hyperfinite probability space, $\mathfrak{A} \subseteq {}^* \mathfrak{P}(V)$ an internal subalgebra generated by a hyperfinite partition $\{V_1, V_2, \dots, V_H\}$ ($H \in {}^* \mathbb{N}$) of V , $X : V \rightarrow {}^* R$ an S -integrable random variable on $(V, {}^* \mathfrak{P}(V), P)$ and $A \in {}^* \mathfrak{P}(V)$. Then*

$$(i) \sum(X(v)P(v), v \in U) = \sum(E(X/\mathfrak{A})(v)P(v), v \in U) \text{ for } U \in \mathfrak{A}$$

(ii) $E(X/\mathfrak{A})$ is the P -a. s. unique internal random variable on (V, \mathfrak{A}, P) which satisfies (i), i.e. for any other S -integrable $Y : V \rightarrow {}^* R$ on (V, \mathfrak{A}, P) satisfying (i)

$$Y(v) \approx E(X/\mathfrak{A})(v) \quad P\text{-n. s.}$$

and for any S -integrable $H : V \rightarrow {}^* R$ on $V, \mathfrak{A}, P)$ with

$$\begin{aligned} \sum(|(H(v) - E(X/\mathfrak{A})(v)|P(v), v \in V) \approx 0 \quad \text{one has} \\ \sum(H(v)P(v), v \in U) \approx \sum(E(X/\mathfrak{A})(v)P(v), v \in U) \quad \text{for } U \in \mathfrak{A}. \end{aligned}$$

(iii) For any set $B \in \mathfrak{A}$

$$P(A \cap B) = \sum(P(A/\mathfrak{A})(v)P(v), v \in B).$$

(iv) $P(A/\mathfrak{A})$ is the P -a. s. unique internal random variable on (V, \mathfrak{A}, P) in the sense given in (ii).

Proof (i) Since $U = \bigcup_{i=1}^H (U \cap V_i)$, for $v \in U \cap V_i$ $E(X/\mathfrak{A})(v) = E(X/U \cap V_i)$

one has

$$\begin{aligned}
\sum (E(X/\mathfrak{A})(v)P(v), v \in U) &= \sum_{i=1}^H (E(X/\mathfrak{A})(v)P(v), v \in \bigcup_{i=1}^H (U \cap V_i)) \\
&= \sum_{i=1}^H (E(X/\mathfrak{A})(v)P(v), v \in U \cap V_i) \\
&= \sum_{i=1}^H (E(X/U \cap V_i)P(U \cap V_i)) \\
&= \sum_{i=1}^H ((P(U \cap V_i))^{-1} \sum (X(u)P(u), u \in U \cap V_i))P(U \cap V_i) \\
&= \sum (X(u)P(u), u \in \bigcup_{i=1}^H (U \cap V_i)) \\
&= \sum (X(u)P(u), u \in U)
\end{aligned}$$

(ii) Let $F(v) = E(X/\mathfrak{A})(v)$. Then, in view of the Projection Theorem for Integrability, [1], S - \mathfrak{A} -integrability of $F : V \rightarrow^* R$ implies that $\sim F : V \rightarrow R$ is a $\sim P$ -integrable random variable on $(V, \mathfrak{M}(\mathfrak{A}), \sim P)$. If $Y : V \rightarrow^* R$ is any S - \mathfrak{A} -integrable random variable on (V, \mathfrak{A}, P) which satisfies (i) then $\sim Y : V \rightarrow R$ is a $\sim P$ -integrable random variable on $(V, \mathfrak{M}(\mathfrak{A}), \sim P)$ as well. Therefore, for $U \in \mathfrak{A}$

$$\int_U \sim Y d\sim P = \text{st}(\sum (Y(u)P(u), u \in U)) = \text{st}(E(F(u)P(u), u \in U)) = \int_U \sim F d\sim P$$

Let $M \in \mathfrak{M}(\mathfrak{A})$. Then, by [8], there exists a set $U \in \mathfrak{A}$ such that $\sim P(U \Delta M) = 0$. Since U satisfies (1), we have

$$\int_M \sim Y d\sim P = \int_U \sim Y d\sim P = \int_U \sim F d\sim P = \int_M \sim F d\sim P$$

and hence $\sim Y(v) = \sim F(v)$ P -a. s.. This implies that

$$P\{v \in V \mid |Y(v) - F(v)| > n^{-1}\} \approx 0 \quad \text{for every } n \in N.$$

According to [4, Robinson's lemma about sequences] there exists $h \in^* NN$ such that for every $k \in^* N$, $k \leq h$

$$P\{v \in V \mid |Y(v) - F(v)| > k^{-1}\} \approx 0$$

Therefore, the set $U = \{v \in V \mid |Y(v) - F(v)| > h^{-1}\} \approx 0$ satisfies: $U \in \mathfrak{A}$, $P(U) \approx 0$, $U \supset \{v \in V \mid Y(v) \neq F(v)\}$ and $Y(v) \approx E(X/\mathfrak{A})(v)$ for $u \notin U$. Hence

$$Y(v) \approx E(X/\mathfrak{A})(v) \quad P\text{-a. s.}$$

Let $H : V \rightarrow {}^*R$ be an S - \mathfrak{A} -integrable random variable on (V, \mathfrak{A}, P) with $\sum(|H(v) - F(v)|P(v), v \in V) \approx 0$. Since $\sim H, \sim F$ are $\sim P$ -integrable random variables on $(V, \mathfrak{M}(\mathfrak{A}), \sim P)$ and

$$\int_V |\sim H - \sim F| d\sim P = \text{st}(\sum(|H(v) - F(v)|P(v), v \in V)) = 0$$

we have that $\sim H(v) = \sim F(v)$ P -a. s.. Therefore, for $U \in \mathfrak{A}$

$$\sum(H(v)P(v), v \in U) \approx \int_U \sim H d\sim P = \int_U \sim F d\sim P \approx \sum(F(v)P(v), v \in U) \quad \text{i.e.}$$

$$\sum(H(v)P(v), v \in U) \approx \sum(E(X/\mathfrak{A})(v)P(v), v \in U)$$

(iii) According to def (2) for $P(A/\mathfrak{A})$ and (i), for $B \in \mathfrak{A}$ we have

$$\begin{aligned} \sum(P(A/\mathfrak{A})(v)P(v), v \in B) &= \sum(E(I_A/\mathfrak{A})(v)P(v), v \in B) \\ &= \sum(I_A(v)P(v), v \in B) = P(A \cap B) \end{aligned}$$

(iv) Let $F(v) = P(A/\mathfrak{A})(v)$. Then, since F is S - \mathfrak{A} -integrable, $\sim F : V \rightarrow R$ is a $\sim P$ -integrable random variable on $(V, \mathfrak{M}(\mathfrak{A}), \sim P)$, so, for an S - \mathfrak{A} -integrable random variable $G : V \rightarrow {}^*R$ on (V, \mathfrak{A}, P) which satisfies (iii), we have that for $B \in \mathfrak{A}$

$$\int_B \sim G d\sim P = \text{st}(\sum(G(v)P(v), v \in B)) = \text{st}(\sum(F(v)P(v), v \in B)) = \int_B \sim F d\sim P$$

Hence, by the same arguments as in proof of (ii)

$$G(v) \approx F(v) = P(A/\mathfrak{A})(v) \quad P\text{-n. s.}$$

For $H : V \rightarrow {}^*R$ which is S - \mathfrak{A} -integrable and satisfies $\sum(|H(v) - F(v)|P(v), v \in V) \approx 0$, like in (ii), we have that $\sim H(v) = \sim F(v)$ P -a. s. implies that for any $B \in \mathfrak{A}$

$$\begin{aligned} \text{st}(\sum(H(v)P(v), v \in B)) &= \int_B \sim H d\sim P = \int_B \sim F d\sim P \\ &= \text{st}(\sum(F(v)P(v), v \in B)) = \text{st}(P(A \cap B)) \end{aligned}$$

$$\text{i.e.} \quad \sum(H(v)P(v), v \in B) \approx P(A \cap B)$$

In [10] it is proved that for S -integrable random variable $X : V \rightarrow {}^*R$ on $(V, {}^*\mathfrak{B}(V), P)$ the internal conditional expectation $E(X/\mathfrak{A})$ is a lifting of $\sim E(\sim X/\mathfrak{M}(\mathfrak{A}))$, $\sim E(\sim X/\mathfrak{M}(\mathfrak{A}))$ being the conditional expectation of $\sim X : V \rightarrow R$ relative to sub- σ -algebra $\mathfrak{M}(\mathfrak{A}) \subseteq \mathfrak{M}(\sim P)$. From this result we derive the following nonstandard characterization of the conditional probability $\sim P(A/\mathfrak{M}(\mathfrak{A}))$, $A \in \mathfrak{M}(\sim P)$ on a Loeb space.

THEOREM 5. Let $(V, * \mathfrak{P}(V), P)$ be a hyperfinite probability space, \mathfrak{A} an internal subalgebra of $* \mathfrak{P}(V)$ and $\sim P(A/\mathfrak{M}(\mathfrak{A}))$ the conditional probability of $A \in \mathfrak{M}(\sim P)$ relative to sub- σ -algebra $\mathfrak{M}(\mathfrak{A}) \subseteq \mathfrak{M}(\sim P)$. Then there exists a set $B \in * \mathfrak{P}(V)$ such that

$$\text{st}(P(B/\mathfrak{A})) = \sim P(A/\mathfrak{M}(\mathfrak{A})) \quad P\text{-a. s.}$$

Proof. For $A \in \mathfrak{M}(\sim P)$ there is a set $B \in * \mathfrak{P}(V)$ such that $\sim P(A \Delta B) = 0$. The indicator function I_B

$$I_B(v) = \begin{cases} 1, v \in B \\ 0, v \notin B \end{cases}$$

is an internal, S -integrable random variable on $(V, * \mathfrak{P}(v), P)$. Since

$$P\{v \mid I_A(v) \neq I_B(v)\} = P(A \Delta B) = 0$$

I_B is an S -integrable lifting of I_A . In view of [10], this implies

$$\sim E(I_B/\mathfrak{A}) = \sim E(I_A/\mathfrak{M}(\mathfrak{A})) \quad P\text{-a. s.} \quad \text{and so}$$

$$\text{st}(P(B/\mathfrak{A})) = \sim P(A/\mathfrak{M}(\mathfrak{A})) \quad P\text{-a. s.}$$

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