

Spectral Orders, Conditional Expectations and Martingales

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1. Introduction

In this paper, we characterize conditional expectations and martingales in terms of the strong spectral order $<$ and submartingales in terms of the weak spectral order \ll . Using these characterizations, we prove that the conditional form of Jensen's inequality is, in fact, a particular case of an extended form of a theorem of Hardy-Littlewood-Pólya obtained in [3, Theorem 2.5], and we give conditions for equality which we have not found mentioned in the literature before. Moreover, we also obtain a new proof of Doob's optional sampling theorem.

If X and Y are integrable random variables defined on a probability space (Ω, \mathcal{F}, P) , then we write $X \ll Y$ whenever

$$E[(X-t)^+] \leq E[(Y-t)^+] \tag{1}$$

for all $t \in \mathbb{R}$, and we write $X < Y$ to mean $X \ll Y$ and $E[X] = E[Y]$.

The notation $<$ and \ll are respectively referred to as the *strong* and *weak spectral order*.

If $X_i < Y_i$ (respectively $X_i \ll Y_i$) and $P[X_i \neq 0, X_j \neq 0] = 0$ and $P[Y_i \neq 0, Y_j \neq 0] = 0$, $1 \leq i < j \leq n$, then it is easy to see that

$$\sum_{i=1}^n X_i < \sum_{i=1}^n Y_i \tag{2}$$

(respectively $\sum_{i=1}^n X_i \ll \sum_{i=1}^n Y_i$). Moreover, if $X < Y$ (respectively $X \ll Y$) and if A is an event containing $\{\omega: X(\omega) \neq Y(\omega)\}$, then

$$X 1_A < Y 1_A \tag{3}$$

(respectively $X 1_A \ll Y 1_A$).

2. Spectral Orders and Conditional Expectations

Theorem 1. *If $X \in L^1(\Omega, \mathcal{F}, P)$ is any random variable, if \mathcal{G} is any sub- σ -algebra of \mathcal{F} , and Y a \mathcal{G} -measurable random variable, then $Y \leq E(X|\mathcal{G})$ (respectively $Y = E(X|\mathcal{G})$) if and only if*

$$Y 1_A \ll X 1_A \quad (Y 1_A < X 1_A) \tag{4}$$

for all $A \in \mathcal{G}$.

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Proof. The condition is clearly sufficient.

To prove its necessity, suppose $Y \leq E(X|\mathcal{G})$. Then $E[Y1_A] \leq E[X1_A]$ for all $A \in \mathcal{G}$. Since Y is \mathcal{G} -measurable, the set $A \cap \{Y1_A > t\}$ belongs to \mathcal{G} for any $A \in \mathcal{G}$ and any $t \in \mathbb{R}$. Thus, $E[Y1_{A \cap \{Y1_A > t\}}] \leq E[X1_{A \cap \{Y1_A > t\}}]$ which implies that $E[(Y1_A - t) 1_{\{Y1_A > t\}}] \leq E[(X1_A - t) 1_{\{Y1_A > t\}}] \leq E[(X1_A - t)^+]$, i.e., $E[(Y1_A - t)^+] \leq E[(X1_A - t)^+]$ for all $A \in \mathcal{G}$ and for all $t \in \mathbb{R}$. Hence (4) follows.

The following corollaries are obtained by the usual standard argument.

Corollary 1. $Y = E(X|\mathcal{G})$ if and only if $YZ < XZ$ for all \mathcal{G} -measurable random variables Z such that $XZ \in L^1(\Omega, \mathcal{F}, P)$.

Corollary 2. $E(XZ|\mathcal{G}) = ZE(X|\mathcal{G})$ for all \mathcal{G} -measurable random variables Z such that $XZ \in L^1(\Omega, \mathcal{F}, P)$.

Theorem 2. (Jensen's Inequality). Suppose $X \in L^1(\Omega, \mathcal{F}, P)$. Then

$$\Phi \circ E(X|\mathcal{G}) \leq E(\Phi \circ X|\mathcal{G}) \quad \text{a.s.} \tag{5}$$

for all convex functions $\Phi: \mathbb{R} \rightarrow \mathbb{R}$ such that $\Phi \circ X \in L^1(\Omega, \mathcal{F}, P)$.

If, in addition, Φ is strictly convex, then equality holds in (5) if and only if $E(X|\mathcal{G})1_A$ and $X1_A$ are identically distributed for each $A \in \mathcal{G}$ or, equivalently, $E(X|\mathcal{G})$ and X are identically distributed.

Proof. Let $Y = E(X|\mathcal{G})$. Since $Y1_A < X1_A$ for all $A \in \mathcal{G}$, we have, by Theorem 2.5 in [3], $E[\Phi(Y1_A)] \leq E[\Phi(X1_A)]$ which implies that

$$\int_A \Phi \circ Y dP \leq \int_A \Phi \circ X dP = \int_A E(\Phi \circ X|\mathcal{G}) dP \quad \text{for all } A \in \mathcal{G}$$

and for all convex functions $\Phi: \mathbb{R} \rightarrow \mathbb{R}$ such that $\Phi \circ X \in L^1(\Omega, \mathcal{F}, P)$. Since both $\Phi \circ Y$ and $E(\Phi \circ X|\mathcal{G})$ are \mathcal{G} -measurable, inequality (5) follows directly from the last inequality.

If Φ is strictly convex such that $\Phi \circ X$ is integrable, then equality holds in (5) if and only if, for all $A \in \mathcal{G}$, $E[\Phi \circ (E(X|\mathcal{G})1_A)] = E[\Phi \circ (X1_A)]$ which is the case if and only if $E(X|\mathcal{G})1_A$ and $X1_A$ are identically distributed, by Theorem 2.3 in [3].

Finally, using the fact that two integrable random variables Y and Z are identically distributed if and only if $Y < Z$ and $Z < Y$, it is easily seen that $E(X|\mathcal{G})$ and X are identically distributed if and only if $E(X|\mathcal{G})1_A$ and $X1_A$ are identically distributed for each $A \in \mathcal{G}$.

3. Spectral Orders and Martingales

In what follows, we let N denote the set of natural numbers, i.e., $N = \{1, 2, 3, \dots\}$, and $\{\mathcal{F}_n: n \in N\}$ an increasing sequence of sub- σ -algebras of \mathcal{F} .

Theorem 3. Suppose $\{X_n, \mathcal{F}_n: n \in N\}$ is a martingale (respectively a submartingale, a supermartingale). Let \mathcal{F}_∞ be the σ -algebra generated by $\mathcal{F}_n, n \in N$. Let $\sigma: \Omega \rightarrow N \cup \{0\}$ and $\tau: \Omega \rightarrow N \cup \{0\}$ be two \mathcal{F}_∞ -measurable random variables such that $\sigma \leq \tau$, $\{\sigma = 0\} = \{\tau = 0\}$, $\{\sigma = n\} \in \mathcal{F}_n$ and $\{\tau = n\} \in \mathcal{F}_n$ for all $n \in N$. Suppose further that one of the following two conditions is satisfied:

- (i) $\{X_n: n \in N\}$ is uniformly integrable;
- (ii) σ and τ are essentially bounded.

If X_0 is any \mathcal{F}_∞ -measurable integrable random variable, then $X_\sigma < X_\tau$ (respectively $X_\sigma \ll X_\tau$, $-X_\sigma \ll -X_\tau$).

Proof. Assume that $\{X_n, \mathcal{F}_n: n \in N\}$ is a martingale.

Let $A_n = \{\sigma = n\}$ and $B_n = \{\tau = n\}$, $n \in N \cup \{0\}$. Then the inequality $\sigma \leq \tau$ implies that $\bigcup_{i=0}^n B_i \subseteq \bigcup_{i=0}^n A_i$ for all $n \in N$. Define $Y_n = \sum_{i=0}^n X_i 1_{B_i} + X_n 1_{C_n} + X_\sigma 1_{D_n}$, where $C_n = \bigcup_{i=0}^n A_i - \bigcup_{i=0}^n B_i$ and $D_n = \Omega - \bigcup_{i=0}^n A_i$, $n \in N$. Clearly $Y_1 = X_\sigma$. We claim that $Y_n < Y_{n+1}$ for all $n \in N$. Now, for each $n \in N$, we have $C_n \in \mathcal{F}_n$ and so $X_n 1_{C_n} < X_{n+1} 1_{C_n}$, by Theorem 2. Thus, it follows from (2) that

$$\begin{aligned} Y_n &= \sum_{i=0}^n X_i 1_{B_i} + X_n 1_{C_n} + X_\sigma 1_{D_n} < \sum_{i=0}^n X_i 1_{B_i} + X_{n+1} 1_{C_n} + X_\sigma 1_{D_n} \\ &= \sum_{i=0}^{n+1} X_i 1_{B_i} + X_{n+1} 1_{C_{n+1}} + X_\sigma 1_{D_{n+1}} = Y_{n+1}, \end{aligned}$$

where $n \in N$.

Now if (i) is satisfied, then it is easily seen that both X_σ and X_τ are integrable, and so $\{Y_n: n \in N\}$ is uniformly integrable, since $|Y_n| \leq |X_\tau| + |X_n| + |X_\sigma|$ for all $n \in N$. Moreover, it is clear that $Y_n \rightarrow X_\tau$ pointwise everywhere. Thus, $Y_n \rightarrow X_\tau$ in L^1 by Theorem T21 in [6, p. 18]. But $X_\sigma = Y_1 < Y_n$ whence we conclude by taking limits that $X_\sigma < X_\tau$.

Next, if (ii) is satisfied, then it is easily seen that there exists a number $m \in N$ such that $Y_n = X_\tau$ a.s. for all $n \geq m$. Hence $X_\sigma < X_\tau$.

The case that $\{X_n, \mathcal{F}_n: n \in N\}$ is a submartingale or a supermartingale is treated analogously.

It is now easy to derive Doob's Optional Sampling Theorem [4, Theorem 2.2, pp. 302-303].

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