

Convergence of Closed Convex Sets and σ -Fields

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Summary. Let $\{C_n\}$ be a sequence of closed convex subsets in a Hilbert space H . We prove that the prediction sequence $\{p(x|C_n)\}$ converges for every $x \in H$ if and only if $s\text{-lim } C_n$ exists and is not empty. We further show the relation between the limit of closed convex sets and the one of σ -subfields in probability measure spaces.

0. Introduction

Let (Ω, Σ, μ) be a probability measure space, and $\{\Sigma_n\}$ be a sequence of σ -subfields of Σ . Now we consider the following proposition:

(E) *There exists a σ -subfield Σ_∞ such that every sequence $\{E(f|\Sigma_n)\}$ of conditional expectations of $f \in L^p(\Omega, \Sigma, \mu)$ converges to $E(f|\Sigma_\infty)$ in L^p -norm ($1 \leq p < \infty$);*

Doob's martingale convergence theorem says that if $\{\Sigma_n\}$ is monotone increasing (resp. decreasing) with respect to set inclusion, Proposition (E) is true for $\Sigma_\infty = \bigvee_n \Sigma_n$ (resp. $\bigcap_n \Sigma_n$). Let $\liminf \Sigma_n$ be the lower limit and $\limsup \Sigma_n$ be the upper limit of $\{\Sigma_n\}$ with respect to set inclusion. Then the proposition is also true, if $\liminf \Sigma_n = \limsup \Sigma_n = \Sigma_\infty$ (see Dang-Ngoc [4]). On the other hand, Neveu [8, IV.3.2] introduced the notion of strong convergence of σ -subfields. We say Σ_∞ to be the strong limit of $\{\Sigma_n\}$ if every sequence $\{P(A|\Sigma_n)\}$ of conditional probabilities of $A \in \Sigma$ converges to $P(A|\Sigma_\infty)$ in measure. Proposition (E) is satisfied if and only if Σ_∞ is the strong limit of $\{\Sigma_n\}$. Becker [2] pointed out that the proposition is the consequence of weak convergence. Kudō [6] studied the notion of strong convergence of σ -subfields more precisely and applied it to the asymptotic theory of statistics. He defined another lower limit $\mu\text{-liminf } \Sigma_n$ and upper limit $\mu\text{-limsup } \Sigma_n$ of $\{\Sigma_n\}$. Σ_∞ is the strong limit of $\{\Sigma_n\}$ if and only if $\mu\text{-liminf } \Sigma_n = \mu\text{-limsup } \Sigma_n = \Sigma_\infty$.

In this paper we investigate analogues theorems in Hilbert spaces. The conditional expectation $E(f|\Sigma')$ of $f \in L^2(\Omega, \Sigma, \mu)$ relative to Σ' is the best ap-

proximation of f by elements of $L^2(\Omega, \Sigma', \mu)$. Hence Proposition (E) in the case of $p=2$ is generalized as follows. Let H be a Hilbert space, and $\{C_n\}$ be a sequence of, non-empty, closed convex subsets of H . Then we consider the following proposition:

(P) *There exists a closed convex subset C_∞ of H such that every sequence $\{p(x|C_n)\}$ of best approximations of $x \in H$ converges to $p(x|C_\infty)$;*

Brunk [3] showed that if $\{C_n\}$ is monotone increasing (resp. decreasing and if $\bigcap_n C_n \neq \emptyset$) with respect to set inclusion, Proposition (P) is true for $C_\infty = \bigcup_n \overline{C_n}$ (resp. $\bigcap_n C_n$), and applied it to the problem of maximum likelihood estimation.

We generalize it for neither increasing nor decreasing case. Mosco [7] defined the strong lower limit $s\text{-liminf } C_n$ and the weak upper limit $w\text{-limsup } C_n$. In Chap.1 we prove that if $s\text{-liminf } C_n = w\text{-limsup } C_n = C_\infty \neq \emptyset$, Proposition (P) is satisfied, and that it is also the consequence of weak convergence. In Chap.2 we study the strong lower limit and the weak upper limit of a sequence of subspaces of H , and in Chap.3 we see the relation between the limit of closed convex sets and the limit of σ -fields. The lower limit and the upper limit of a sequence of σ -subfields are corresponding to the strong lower limit and the weak upper limit of the sequence of subspaces defined by the σ -subfields in $L^2(\Omega, \Sigma, \mu)$ respectively.

1. The Limit of a Sequence of Closed Convex Sets

Let H be a Hilbert space with inner product $\langle \cdot | \cdot \rangle$ and norm $\| \cdot \|$, and \mathfrak{C} be the set of all, non-empty, closed convex subsets of H . For any $x \in H$ and $C \in \mathfrak{C}$ there exists a unique closest point $p(x|C)$ of C to x . The following lemma is well known (see, for example, Brunk [3]).

Lemma 1.1. *Let $C \in \mathfrak{C}$ and $x, y \in H$. Then*

- (i) $x' = p(x|C)$ if and only if $x' \in C$ and $\text{Re} \langle x - x' | x' - z \rangle \geq 0$ for any $z \in C$;
- (ii) $\|p(x|C) - p(y|C)\| \leq \|x - y\|$; That is, $p(\cdot|C)$ is non-expansive.

Let $\{C_n\}$ be a sequence in \mathfrak{C} . We can define a lower limit $\text{liminf } C_n$ and an upper limit $\text{limsup } C_n$ of $\{C_n\}$ with respect to set inclusion as

$$\text{liminf } C_n = \bigcup_{m=1}^{\infty} \bigcap_{n=m}^{\infty} C_n, \quad \text{limsup } C_n = \bigcap_{m=1}^{\infty} \overline{\bigcup_{n=m}^{\infty} C_n},$$

where $\overline{}$ means the closed convex hull. On the other hand, Mosco [7] defined a strong lower limit and a weak upper limit of a sequence of subsets of Banach spaces. Following it, we define a strong lower limit $s\text{-liminf } C_n$ and a weak upper limit $w\text{-limsup } C_n$ of $\{C_n\}$ as

$$\begin{aligned} s\text{-liminf } C_n &= \{x \in H: x_n \rightarrow x \text{ as } n \rightarrow \infty, x_n \in C_n \text{ for every } n\}, \\ w\text{-limsup } C_n &= \overline{\{x \in H: x_{n'} \rightarrow x \text{ (weakly) as } n' \rightarrow \infty, x_{n'} \in C_{n'} \\ &\quad \text{for every } n', \{C_{n'}\} \text{ is a subsequence of } \{C_n\}\}}. \end{aligned}$$

If $s\text{-liminf } C_n = w\text{-limsup } C_n$, then the common value is denoted by $s\text{-lim } C_n$. Some elementary properties and examples are seen in [7].

Proposition 1.2. *Let $\{C_n\}$ be a sequence in \mathfrak{C} .*

- (i) $s\text{-liminf } C_n = \{x \in H : p(x|C_n) \rightarrow x \text{ as } n \rightarrow \infty\}$;
- (ii) $s\text{-liminf } C_n$ and $w\text{-limsup } C_n$ are in $\mathfrak{C} \cup \{\emptyset\}$;
- (iii) $\text{liminf } C_n \subset s\text{-liminf } C_n \subset w\text{-limsup } C_n \subset \text{limsup } C_n$.

Proof. (i): If $x \in s\text{-liminf } C_n$, then

$$\|x - p(x|C_n)\| \leq \|x - x_n\| \rightarrow 0,$$

as $n \rightarrow \infty$. The converse is trivial.

(ii): Let $x \in s\text{-liminf } C_n$. Then for any $\varepsilon > 0$ there exists $x_0 \in s\text{-liminf } C_n$ with $\|x - x_0\| < \varepsilon/3$. On the other hand, by (i) we have an n_0 such that $\|x_0 - p(x_0|C_n)\| < \varepsilon/3$ for any $n \geq n_0$. Hence for $n \geq n_0$

$$\begin{aligned} \|x - p(x|C_n)\| &\leq \|x - x_0\| + \|x_0 - p(x_0|C_n)\| \\ &\quad + \|p(x_0|C_n) - p(x|C_n)\| \\ &\leq 2 \cdot \|x - x_0\| + \|x_0 - p(x_0|C_n)\| \leq \varepsilon. \end{aligned}$$

Thus $\|x - p(x|C_n)\| \rightarrow 0$ as $n \rightarrow \infty$, and we have $x \in s\text{-liminf } C_n$. Therefore $s\text{-liminf } C_n$ is closed. It is clear from the definition that $s\text{-liminf } C_n$ is convex and $w\text{-limsup } C_n$ is closed convex. If $C_n = \{x_n\}$ and $\|x_n\| \rightarrow \infty$ as $n \rightarrow \infty$, then $s\text{-liminf } C_n = w\text{-limsup } C_n = \emptyset$.

(iii): The first and second inclusions are clear from the definition. The last one follows from the fact that closed convex sets are weakly closed. \square

Lemma 1.3. *Let $\{C_n\}$ be a sequence in \mathfrak{C} such that $s\text{-liminf } C_n \neq \emptyset$. Then for every $x \in H$*

- (i) $\sup \|p(x|C_n)\| < \infty$;
- (ii) *If $\{p(x|C_n)\}$ weakly converges to x , then $\|x - p(x|C_n)\| \rightarrow 0$ as $n \rightarrow \infty$.*

Proof. Let x_0 be an element of $s\text{-liminf } C_n$. Since $\{p(x_0|C_n)\}$ is a norm converging sequence by Proposition 1.2 (i), $\sup \|p(x_0|C_n)\| < \infty$. On the other hand

$$\begin{aligned} \|p(x|C_n)\| &\leq \|p(x|C_n) - x\| + \|x\| \\ &\leq \|p(x_0|C_n) - x\| + \|x\| \\ &\leq \|p(x_0|C_n)\| + 2 \cdot \|x\|, \end{aligned}$$

for every n . Hence we have (i). If $\{p(x|C_n)\}$ weakly converges to x ,

$$\begin{aligned} \|x - p(x|C_n)\|^2 &\leq \|x - p(x|C_n)\|^2 + \text{Re} \langle x - p(x|C_n) | p(x|C_n) - p(x_0|C_n) \rangle \\ &= \text{Re} \langle x - p(x|C_n) | x - x_0 \rangle + \text{Re} \langle x - p(x|C_n) | x_0 - p(x_0|C_n) \rangle \\ &\leq \text{Re} \langle x - p(x|C_n) | x - x_0 \rangle \\ &\quad + (\|x\| + \|p(x|C_n)\|) \cdot \|x_0 - p(x_0|C_n)\| \rightarrow 0, \end{aligned}$$

as $n \rightarrow \infty$. Thus we have (ii). \square

Theorem 1.4. *Let $\{C_n\}$ be a sequence in \mathfrak{C} . Then the following assertions are equivalent:*

- (i) $s\text{-lim } C_n$ exists and is not empty;
 - (ii) $s\text{-liminf } C_n$ is not empty and there exists $C_\infty \in \mathfrak{C}$ such that $\{p(x|C_n)\}$ weakly converges to $p(x|C_\infty)$ for every $x \in H$;
 - (iii) $\{p(x|C_n)\}$ is a norm convergence sequence for every $x \in H$.
- Moreover, if these assertions are satisfied, $\{p(x|C_n)\}$ converges to $p(x|s\text{-lim } C_n)$ for every $x \in H$.

Proof. (i) \Rightarrow (ii): Let $C_\infty = s\text{-lim } C_n$. For any $x \in H$, by Lemma 1.3 (i), $\{p(x|C_n)\}$ is norm bounded. Hence for any subsequence $\{p(x|C_{n'})\}$ of $\{p(x|C_n)\}$ there exists a subsequence $\{p(x|C_{n''})\}$ which weakly converges to some $x' \in H$. Then $x' \in w\text{-limsup } C_n = C_\infty$. On the other hand, for any $y \in C_\infty$, since $y = s\text{-lim } p(y|C_n)$, we have

$$\|x - x'\| \leq \lim \|x - p(x|C_{n'})\| \leq \lim \|x - p(y|C_{n'})\| = \|x - y\|,$$

where the first inequality follows from weak lower semicontinuity of norm $\|\cdot\|$. Thus $x' = p(x|C_\infty)$ and we have (ii).

(ii) \Rightarrow (iii): Any $x \in H$ is fixed. Since $\{p(p(x|C_\infty)|C_n)\}$ weakly converges to $p(x|C_\infty)$, by Lemma 1.3 (ii), we have $p(x|C_\infty) = s\text{-lim } p(p(x|C_\infty)|C_n)$. Hence by Lemma 1.1 (i)

$$\begin{aligned} \limsup \|p(x|C_n)\|^2 &\leq \limsup \|p(x|C_n)\|^2 \\ &\quad + \liminf \langle x - p(x|C_n) | p(x|C_n) - p(p(x|C_\infty)|C_n) \rangle \\ &= \|p(x|C_\infty)\|^2. \end{aligned}$$

On the other hand, because of weak lower semicontinuity of norm $\|\cdot\|$, we have $\liminf \|p(x|C_n)\| \geq \|p(x|C_\infty)\|$. Thus $\lim \|p(x|C_n)\| = \|p(x|C_\infty)\|$, and by one of the convex properties of Hilbert space norm (see, for example, Day [5]) we have $p(x|C_\infty) = s\text{-lim } p(x|C_n)$.

(iii) \Rightarrow (i): We put $C_\infty = \{s\text{-lim } p(x|C_n) : x \in H\}$. It is clear that $C_\infty \subset s\text{-liminf } C_n$. Now assume that x is an element such that there exist a subsequence $\{C_{n'}\}$ of $\{C_n\}$ and $x_{n'} \in C_{n'}$ for every n' with $x_{n'} \rightarrow x$ (weakly) as $n' \rightarrow \infty$. Let $y = s\text{-lim } p(x|C_n)$. From Lemma 1.1 we have

$$\operatorname{Re} \langle x - p(x|C_{n'}) | p(x|C_{n'}) - x_{n'} \rangle \geq 0$$

for every n' . Tending n' to ∞ , $\operatorname{Re} \langle x - y | y - x \rangle \geq 0$. Therefore $x = y \in C_\infty$. Thus $C_\infty = s\text{-liminf } C_n = w\text{-limsup } C_n$. \square

If $\{C_n\}$ is increasing (resp. decreasing and if $\bigcap_n C_n \neq \emptyset$), then by Proposition 1.2 (iii) and the above theorem it follows that $\{p(x|C_n)\}$ converges to $p\left(x \left| \bigcup_{n=1}^\infty C_n \right.\right)$ (resp. $p\left(x \left| \bigcap_{n=1}^\infty C_n \right.\right)$) for every $x \in H$. See Brunk [3].

2. The Limit of a Sequence of Subspaces

Let $\{H_n\}$ be a sequence of subspaces (i.e., closed linear manifolds) of H . Then we can easily see that both $s\text{-liminf } H_n$ and $w\text{-limsup } H_n$ are subspaces of H .

Proposition 2.1. *Let $\{H_n\}$ be a sequence of subspaces of H . Then*

- (i) $s\text{-liminf } H_n^\perp = (w\text{-limsup } H_n)^\perp$;
- (ii) $w\text{-limsup } H_n^\perp = (s\text{-liminf } H_n)^\perp$.

Proof. For the simplicity we denote $p(\cdot|H_n)$ by P_n , which is the orthogonal projection onto H_n , for every n . Suppose that $x \in s\text{-liminf } H_n^\perp$. Then $\{P_n x\}$ converges to 0. Let y be any element such that there exists a subsequence $\{H_{n'}\}$ and $y_{n'} \in H_{n'}$ for every n' with $y_{n'} \rightarrow y$ (weakly) as $n' \rightarrow \infty$. Then

$$\begin{aligned} |\langle x|y \rangle| &= \lim |\langle x|y_{n'} \rangle| = \lim |\langle P_{n'} x|y_{n'} \rangle| \\ &\leq \lim \|P_{n'} x\| \cdot \|y_{n'}\| = 0. \end{aligned}$$

Therefore $x \in (s\text{-liminf } H_n)^\perp$. Conversely suppose that $x \in (w\text{-limsup } H_n)^\perp$. It suffices to prove that $\{P_n x\}$ converges to 0. Since $\{P_n x\}$ is norm bounded, for any subsequence $\{P_{n'} x\}$ of $\{P_n x\}$ there exists a subsubsequence $\{P_{n''} x\}$ which weakly converges to some $y \in H$. Then $y \in w\text{-limsup } H_n$. Hence we have

$$\|P_{n''} x\|^2 = \langle P_{n''} x|x \rangle \rightarrow \langle y|x \rangle = 0,$$

as $n'' \rightarrow \infty$. Therefore $\{P_{n'} x\}$ converges to 0, and $\{P_n x\}$ does. Thus (i) is proved. (ii) follows from (i). \square

Theorem 2.2. *Let $\{H_n\}$ be a sequence of subspaces of H . Then*

- (i) $s\text{-liminf } H_n$ is the maximum subspace among subspaces H' of H with

$$\liminf \|p(x|H_n)\| \geq \|p(x|H')\|, \tag{1}$$

for all $x \in H$;

- (ii) $w\text{-limsup } H_n$ is the minimum subspace among subspaces H' of H with

$$\limsup \|p(x|H_n)\| \leq \|p(x|H')\|, \tag{2}$$

for all $x \in H$.

Proof. We denote $p(\cdot|H')$ (resp. $p(\cdot|H_n)$) by P' (resp. P_n).

(i) We first show that $H' = s\text{-liminf } H_n$ satisfies (1). Since $P_n x \rightarrow x$ as $n \rightarrow \infty$ for every $x \in H'$,

$$\begin{aligned} \|P_n x\|^2 &= \|P_n P' x + P_n(1-P')x\|^2 \\ &= \langle P_n P' x|P' x \rangle + \langle P_n P' x|(1-P')x \rangle \\ &\quad + \langle (1-P')x|P_n P' x \rangle + \|P_n(1-P')x\|^2 \\ &\geq \langle P_n P' x|P' x \rangle + \langle P_n P' x|(1-P')x \rangle \\ &\quad + \langle (1-P')x|P_n P' x \rangle \rightarrow \|P' x\|^2, \end{aligned}$$

as $n \rightarrow \infty$ for any $x \in H$. Conversely we shall show that any subspace H' of H satisfying (1) is contained in $s\text{-liminf } H_n$. It suffices to prove that $\{P_n P' x\}$

converges to $P'x$ for any $x \in H$. From (1) it follows that

$$\|P'x\|^2 \leq \liminf \|P_n P'x\|^2 \leq \limsup \|P_n P'x\|^2 \leq \|P'x\|^2,$$

and that $\lim \|P_n P'x\| = \|P'x\|$. Hence

$$\begin{aligned} \|P'x - P_n P'x\|^2 &= \|P'x\|^2 - \langle P'x | P_n P'x \rangle - \langle P_n P'x | P'x \rangle + \langle P_n P'x | P'x \rangle \\ &= \|P'x\|^2 - \|P_n P'x\|^2 \rightarrow 0, \end{aligned}$$

as $n \rightarrow \infty$.

(ii): Now suppose that $H' = w\text{-}\limsup H_n$. Then $H'^{\perp} = s\text{-}\liminf H_n^{\perp}$ and from (1) we have

$$\liminf \|(1 - P_n)x\|^2 \geq \|(1 - P')x\|^2,$$

for any $x \in H$. Hence

$$\|x\|^2 - \limsup \|P_n x\|^2 \geq \|x\|^2 - \|P'x\|^2,$$

and we have $\limsup \|P_n x\| \leq \|P'x\|$ for any $x \in H$. Conversely let H' be a subspace of H with (2). Then by the converse calculation above we have $H' \supset w\text{-}\limsup H_n$. \square

3. The Limit of a Sequence of σ -fields

Let (Ω, Σ, μ) be a probability measure space. For the simplicity we assume that Σ and every σ -subfield of Σ considered in this section are μ -complete (i.e., all μ -null sets are contained). Given a σ -subfield Σ' , we denote by $L^p(\Sigma')$ the Banach space of p -th integrable Σ' -measurable functions, and by $E(\cdot | \Sigma')$ the conditional expectation with respect to Σ' . The norm on $L^p(\Sigma)$ is denoted by $\|\cdot\|_p$.

Let $\{\Sigma_n\}$ be a sequence of σ -subfields. Kudō [6] defined a *lower limit* $\mu\text{-}\liminf \Sigma_n$ and an *upper limit* $\mu\text{-}\limsup \Sigma_n$ as follows:

(i) $\mu\text{-}\liminf \Sigma_n$ is the σ -subfield Σ_0 such that if $\Sigma' = \Sigma_0$, then for every bounded measurable function f

$$\liminf \|E(f | \Sigma_n)\|_1 \geq \|E(f | \Sigma')\|_1, \tag{1}$$

and that any σ -subfield Σ' satisfying (1) is contained in Σ_0 ;

(ii) $\mu\text{-}\limsup \Sigma_n$ is the σ -subfield Σ_0 such that if $\Sigma' = \Sigma_0$, then for every bounded measurable function f

$$\limsup \|E(f | \Sigma_n)\|_1 \leq \|E(f | \Sigma')\|_1, \tag{2}$$

and that any σ -subfield Σ' satisfying (2) contains Σ_0 .

Theorem 3.1 (Kudō [6]). *Let $\{\Sigma_n\}$ be a sequence of σ -subfields. Then*

$$\begin{aligned} \mu\text{-}\liminf \Sigma_n &= \{A \in \Sigma : \text{there exist } A_n \in \Sigma_n \text{ for every } n \text{ with} \\ &\mu(A_n \Delta A) \rightarrow 0 \text{ as } n \rightarrow \infty\}. \end{aligned}$$

On the other hand, since $\{L^2(\Sigma_n)\}$ is a sequence of subspace of Hilbert space $L^2(\Sigma)$, we can define $s\text{-}\liminf L^2(\Sigma_n)$ and $w\text{-}\limsup L^2(\Sigma_n)$ as is defined in the previous section.

Theorem 3.2. *Let $\{\Sigma_n\}$ be a sequence of σ -subfields. Then $s\text{-}\liminf L^2(\Sigma_n) = L^2(\mu\text{-}\liminf \Sigma_n)$.*

Proof. We first show that there exists a σ -subfield Σ' such that $L^2(\Sigma') = s\text{-}\liminf L^2(\Sigma_n)$. It suffices to prove that $s\text{-}\liminf L^2(\Sigma_n)$ is a lattice and has constant functions (see Schaefer [10, Proposition 11.2]). Let f and g belong to $s\text{-}\liminf L^2(\Sigma_n)$. Then there exist $f_n, g_n \in L^2(\Sigma_n)$ for every n such that $f_n \rightarrow f$ and $g_n \rightarrow g$ as $n \rightarrow \infty$ in the square mean. Since $(a \vee b - c \vee d)^2 \leq (a - c)^2 + (b - d)^2$ for any $a, b, c, d \in \mathbb{R}$, where $a \vee b$ (resp. $c \vee d$) is the maximum of $\{a, b\}$ (resp. $\{c, d\}$), we have that $\|f_n \vee g_n - f \vee g\|_2^2 \leq \|f_n - f\|_2^2 + \|g_n - g\|_2^2 \rightarrow 0$ as $n \rightarrow \infty$. Hence we have $f \vee g \in s\text{-}\liminf L^2(\Sigma_n)$, because $f_n \vee g_n \in L^2(\Sigma_n)$ for every n . It is clear that $s\text{-}\liminf L^2(\Sigma_n)$ has constant functions. Hence we define Σ' as the σ -subfield such as $L^2(\Sigma') = s\text{-}\liminf L^2(\Sigma_n)$. If $\Sigma' = \mu\text{-}\liminf \Sigma_n$ is shown, the theorem is proved.

Let A belong to Σ' . Then there exist $f_n \in L^2(\Sigma_n)$ for every n such that $f_n \rightarrow 1_A$ as $n \rightarrow \infty$ in the square mean and hence in measure. For each n we define $A_n = \{\omega \in \Omega : f_n(\omega) \geq 1/2\}$. Then $\mu(A_n \Delta A) \rightarrow 0$ as $n \rightarrow \infty$, because $A_n \Delta A = \{\omega \in \Omega : (1_A(\omega) = 0 \text{ and } f_n(\omega) \geq 1/2) \text{ or } (1_A(\omega) = 1 \text{ and } f_n(\omega) < 1/2)\} \subset \{\omega \in \Omega : |1_A(\omega) - f_n(\omega)| \geq 1/2\}$ for every n . Since $A_n \in \Sigma_n$ for every n , by Theorem 3.1 we have $A \in \mu\text{-}\liminf \Sigma_n$.

Conversely let $A \in \mu\text{-}\liminf \Sigma_n$. Then by Theorem 3.1 there exist $A_n \in \Sigma_n$ for every n such that $\mu(A_n \Delta A) \rightarrow 0$ as $n \rightarrow \infty$. Hence

$$\|1_{A_n} - 1_A\|_2^2 = \int |1_{A_n} - 1_A|^2 d\mu = \int 1_{A_n \Delta A} d\mu = \mu(A_n \Delta A) \rightarrow 0,$$

as $n \rightarrow \infty$. Thus $1_A \in s\text{-}\liminf L^2(\Sigma_n)$ and $A \in \Sigma'$. \square

Theorem 3.3. *Let $\{\Sigma_n\}$ be a sequence of σ -subfields. Then $\mu\text{-}\liminf \Sigma_n$ is the maximum σ -subfield among σ -subfields Σ' of Σ with*

$$\liminf \|E(f|\Sigma_n)\|_2 \geq \|E(f|\Sigma')\|_2, \tag{3}$$

for every $f \in L^2(\Sigma)$.

Proof. Since $E(\cdot|\Sigma_n)$ is the orthogonal projection onto $L^2(\Sigma_n)$ for every n and from Theorem 3.2 $E(\cdot|\mu\text{-}\liminf \Sigma_n)$ is the one onto $s\text{-}\liminf L^2(\Sigma_n)$, it follows from Theorem 2.2 (i) that $\Sigma' = \mu\text{-}\liminf \Sigma_n$ satisfies (3).

Conversely let Σ' be a σ -subfield satisfying (3). Then by Theorem 2.2 (i) and Theorem 3.2 we have $L^2(\mu\text{-}\liminf \Sigma_n) \supset L^2(\Sigma')$. Hence $\mu\text{-}\liminf \Sigma_n \supset \Sigma'$. \square

Theorem 3.4. *Let $\{\Sigma_n\}$ be a sequence of σ -subfields. Then for any $f \in L^1(\Sigma)$ the following assertions are equivalent:*

- (i) f is $\mu\text{-}\liminf \Sigma_n$ -measurable;
- (ii) $\|E(f|\Sigma_n) - f\|_1 \rightarrow 0$ as $n \rightarrow \infty$.

Proof. (i) \Rightarrow (ii): Since $f \in L^1(\mu\text{-}\liminf \Sigma_n)$, for any $\varepsilon > 0$ there exists $f_0 \in L^2(\mu\text{-}\liminf \Sigma_n)$ with $\|f - f_0\|_1 \leq \varepsilon/3$. By Theorem 3.2 and Proposition 1.2 (i) we have that $\|E(f_0|\Sigma_n) - f_0\|_2 \rightarrow 0$ as $n \rightarrow \infty$. Therefore there exists n_0 such that for any $n \geq n_0$

$$\begin{aligned}
\|E(f|\Sigma_n) - f\|_1 &\leq \|E(f|\Sigma_n) - E(f_0|\Sigma_n)\|_1 \\
&\quad + \|E(f_0|\Sigma_n) - f_0\|_1 + \|f_0 - f\|_1 \\
&\leq 2 \cdot \|f - f_0\|_1 + \|E(f_0|\Sigma_n) - f_0\|_2 < \varepsilon.
\end{aligned}$$

Thus $\|E(f|\Sigma_n) - f\|_1 \rightarrow 0$ as $n \rightarrow \infty$.

(ii) \Rightarrow (i): We define $M = \{f \in L^1(\Sigma) : \|E(f|\Sigma_n) - f\|_1 \rightarrow 0 \text{ as } n \rightarrow \infty\}$. Let $f, g \in M$. Then in the same way of the proof of Theorem 3.2 we have $\|E(f|\Sigma_n) \vee E(g|\Sigma_n) - f \vee g\|_1 \rightarrow 0$ as $n \rightarrow \infty$. Therefore

$$\begin{aligned}
\|E(f \vee g|\Sigma_n) - f \vee g\|_1 &\leq \|E(f \vee g|\Sigma_n) - E(f|\Sigma_n) \vee E(g|\Sigma_n)\|_1 \\
&\quad + \|E(f|\Sigma_n) \vee E(g|\Sigma_n) - f \vee g\|_1 \\
&= \|E(f \vee g - E(f|\Sigma_n) \vee E(g|\Sigma_n)|\Sigma_n)\|_1 \\
&\quad + \|E(f|\Sigma_n) \vee E(g|\Sigma_n) - f \vee g\|_1 \\
&\leq 2 \cdot \|f \vee g - E(f|\Sigma_n) \vee E(g|\Sigma_n)\|_1 \rightarrow 0,
\end{aligned}$$

as $n \rightarrow \infty$. Thus $f \vee g \in M$. It can be easily seen that M is closed linear and has constant functions. Hence there exists a σ -subfield Σ' such that $M = L^1(\Sigma')$. For any $f \in L^\infty(\Sigma')$

$$\begin{aligned}
\|E(f|\Sigma_n) - f\|_2^2 &\leq \|E(f|\Sigma_n) - f\|_\infty \cdot \|E(f|\Sigma_n) - f\|_1 \\
&\leq 2 \cdot \|f\|_\infty \cdot \|E(f|\Sigma_n) - f\|_1 \rightarrow 0,
\end{aligned}$$

as $n \rightarrow \infty$. Therefore $f \in L^2(\mu\text{-liminf } \Sigma_n)$. Thus $L^\infty(\Sigma') \subset L^2(\mu\text{-liminf } \Sigma_n)$. Taking here the closure in L^1 -norm for both sides, we have $M \subset L^1(\mu\text{-liminf } \Sigma_n)$. \square

Let $\{\Sigma_n\}$ be a sequence of σ -subfields of Σ . We define $(\Sigma)_p$ as the family of σ -subfields Σ' of Σ with

$$\limsup \|E(f|\Sigma_n)\|_p \leq \|E(f|\Sigma')\|_p,$$

for every bounded measurable function f ($p = 1, 2$).

Lemma 3.5. *If $\Sigma' \in (\Sigma)_1 \cup (\Sigma)_2$, then*

$$\limsup \|E(E(f|\Sigma')|\Sigma_n)\|_p = \limsup \|E(f|\Sigma_n)\|_p,$$

for every bounded measurable function f and $p = 1, 2$.

Proof. Let $\Sigma' \in (\Sigma)_1$. Then for every bounded measurable function f we have

$$\limsup \|E(E(f|\Sigma') - f|\Sigma_n)\|_1 \leq \|E(E(f|\Sigma') - f|\Sigma')\|_1 = 0.$$

Hence

$$\begin{aligned}
&|\limsup \|E(E(f|\Sigma')|\Sigma_n)\|_1 - \limsup \|E(f|\Sigma_n)\|_1| \\
&\leq \limsup \|E(E(f|\Sigma')|\Sigma_n) - E(f|\Sigma_n)\|_1 = 0.
\end{aligned}$$

On the other hand, for every bounded measurable function f

$$\begin{aligned}
&\limsup \|E(E(f|\Sigma') - f|\Sigma_n)\|_2^2 \\
&\leq \limsup \|E(E(f|\Sigma') - f|\Sigma_n)\|_\infty \cdot \|E(E(f|\Sigma') - f|\Sigma_n)\|_1 \\
&\leq \|E(f|\Sigma') - f\|_\infty \cdot \limsup \|E(E(f|\Sigma') - f|\Sigma_n)\|_1 = 0.
\end{aligned}$$

Hence

$$\begin{aligned} &|\limsup \|E(E(f|\Sigma')|\Sigma_n)\|_2 - \limsup \|E(f|\Sigma_n)\|_2| \\ &\leq \limsup \|E(E(f|\Sigma')|\Sigma_n) - E(f|\Sigma_n)\|_2 = 0. \end{aligned}$$

The lemma is proved for $\Sigma' \in (\Sigma)_2$ similarly. \square

Lemma 3.6. *If $\Sigma' \in (\Sigma)_1$ and $\Sigma'' \in (\Sigma)_2$, then $\Sigma' \cap \Sigma'' \in (\Sigma)_1 \cap (\Sigma)_2$.*

Proof. We denote $E(\cdot|\Sigma')$ and $E(\cdot|\Sigma'')$ by E' and E'' respectively. Let f be a bounded measurable function. We show that $\limsup \|E(f|\Sigma_n)\|_1 \leq \|(E' E'')^m f\|_1$ for every $m \in \mathbb{N}$. This is true for $m=1$. We assume that it is true for $m=k$. Then using Lemma 3.5 twice, we have that

$$\begin{aligned} \|(E' E'')^{k+1} f\|_1 &= \|(E' E'')^k E' E'' f\|_1 \geq \limsup \|E(E' E'' f|\Sigma_n)\|_1 \\ &= \limsup \|E(f|\Sigma_n)\|_1. \end{aligned}$$

Hence it is true for $m=k+1$. Tending here $m \rightarrow \infty$, we have

$$\limsup \|E(f|\Sigma_n)\|_1 \leq \|E(f|\Sigma' \cap \Sigma'')\|_1$$

(see [6, Lemma 3.2]). We can prove this inequality for norm $\|\cdot\|_2$ in the same way. Thus we have the lemma. \square

Theorem 3.7. *Let $\{\Sigma_n\}$ be a sequence of σ -subfields. Then μ - $\limsup \Sigma_n$ is the minimum σ -subfield among σ -subfields Σ' of Σ with*

$$\limsup \|E(f|\Sigma_n)\|_2 \leq \|E(f|\Sigma')\|_2, \tag{4}$$

for every $f \in L^2(\Sigma)$.

Proof. We can easily see that (4) is satisfied for every $f \in L^2(\Sigma)$ if and only if it is so for every bounded measurable function f . On the other hand, by Lemma 3.6 we have $(\Sigma)_1 = (\Sigma)_2$. Hence the theorem is proved. \square

Theorem 3.8. *Let $\{\Sigma_n\}$ be a sequence of σ -subfields. Then the closed linear lattice generated by w - $\limsup L^2(\Sigma_n)$ is equal to $L^2(\mu$ - $\limsup \Sigma_n)$.*

Proof. It is the direct consequence of Theorem 2.2(ii) and Theorem 3.7. \square

We discuss in [13] the relations between the results of Sect. 1 and 2 and the geometrical properties of Banach space norms. On the other hand, a referee pointed out in the light of his unpublished works that the theorems in Sect. 3 hold on general L^p spaces ($1 \leq p < \infty$). The author would like to express his gratitude to this referee for his very useful comments and also to Professor H. Umegaki for his valuable advice and constant encouragement.

References

1. Ando, T., Amemiya, I.: Almost everywhere convergence of prediction sequences in L^p . *Z. Wahrscheinlichkeitstheorie verw. Gebiete* **4**, 113–120 (1965)
2. Becker, R.: Ensembles compacts de tribus. *Z. Wahrscheinlichkeitstheorie verw. Gebiete* **29**, 229–234 (1974)

3. Brunk, H.D.: Conditional expectation given a σ -lattice and applications. *Ann. Math. Statist.* **36**, 1339–1350 (1965)
4. Dang-Ngoc, N.: Convergence forte des espérances conditionnelles et des projecteurs d'un espace de Hilbert. *Ann. Inst. H. Poincaré, Sect. B*, VI, 9–13 (1970)
5. Day, M.M.: *Normed Linear Spaces*, 3rd ed. Berlin-Heidelberg-New York: Springer 1973
6. Kudō, H.: A note of the strong convergence of σ -algebras. *Ann. Probab.* **2**, 76–83 (1974)
7. Mosco, U.: Convergence of convex sets and of solutions of variational inequalities. *Adv. Math.* **3**, 510–585 (1969)
8. Neveu, J.: *Bases Mathématiques du Calcul des Probabilités*. Paris: Masson 1964
9. Rao, M.M.: Prediction sequences in smooth Banach spaces. *Ann. Inst. H. Poincaré, Sect. B*, **VIII**, 319–322 (1972)
10. Schaefer, H.H.: *Banach Lattices and Positive Operators*. Berlin-Heidelberg-New York: Springer 1974
11. Shintani, T., Ando, T.: Best approximants in L^1 space. *Z. Wahrscheinlichkeitstheorie verw. Gebiete* **33**, 33–39 (1975)
12. Singer, I.: *Best Approximation in Normed Linear Spaces by Elements of Linear Subspaces*. Berlin-Heidelberg-New York: Springer 1973
13. Tsukada, M.: Convergence of best approximations in a smooth Banach space. preprint (1982)

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