# Prediction Operators in Banach Ideal Spaces ${ }^{1}$ 

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A necessary and sufficient condition for an operator in a Banach ideal space to be a prediction operator is given. © 2001 Elsevier Science (USA)

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Let $X$ be a Banach space, $C$ a set in $X$, and $x \in X$. An element $y \in C$ is called an element of best approximation of $x$ in $C$ if we have

$$
\|x-y\|=\inf \{\|x-z\|: z \in C\}
$$

If the element of best approximation of $x$ in $C$ is unique, we denote it by $\pi(x \mid C)$. The operator $\pi(\cdot \mid C)$ is called the best approximation operator of $X$ onto $C$.

A space $X$ is said to be strictly convex if for each pair $x, y \in X$ with $\|x\|=\|y\|=1$ and $x \neq y$ we have $\|x+y\|<1$. The space $X$ is said to be smooth if at each point $0 \neq x \in X$, there is only one support functional $f_{x}$, i.e., $f_{x} \in X^{*},\left\|f_{x}\right\|_{X^{*}}=1$, and $f_{x}(x)=\|x\|$.

Let $(G, \Sigma, \mu)$ be a $\sigma$-finite measure space and denote by $S=S(G, \Sigma, \mu)$ the set of all equivalence classes of $\Sigma$-measurable real-valued functions on $G$ with algebraic operations and order defined in a natural way. A linear subset $X$ of $S$ endowed with some norm $\|\cdot\|$ is called a Banach ideal space if $(X,\|\cdot\|)$ is complete and it satisfies the condition that $x \in X, y \in S$, and

[^0]$|y| \leqslant|x|$ imply $y \in X$ and $\|y\| \leqslant\|x\|$. The norm on a Banach ideal space is called order-continuous if in addition it satisfies the condition that $x_{n} \downarrow 0$ implies $\left\|x_{n}\right\| \rightarrow 0$. It is well known that $L^{p}(1 \leqslant p \leqslant \infty)$, Orlicz spaces, Lorentz spaces, Orlicz-Lorentz spaces, and Musielak-Orlicz spaces are all Banach ideal spaces. Criterions of order-continuity, strict convexity, smoothness, and reflexivity of these spaces can be found in the literature.

For a Banach ideal space $X(G, \Sigma, \mu)$ (which is simply denoted by $X(\Sigma)$ ), let $\Sigma^{\prime}$ be a $\sigma$-sublattice of the $\sigma$ algebra $\Sigma$ and $X\left(\Sigma^{\prime}\right)=\left\{x \in X(\Sigma): x\right.$ is $\Sigma^{\prime}$ measurable $\}$. If $\pi\left(\cdot \mid X\left(\Sigma^{\prime}\right)\right)$ exists, it is called a prediction operator. Prediction operators have wide applications in probability, Bayes estimation theory, prediction theory, and many other fields. Various authors have studied these in the past 30 years [2-12]. For example, Dykstra has studied the case $L^{2}$ in [4]. Landers and Rogge have studied the case $L^{p}$ in [6] and then obtained a result of Orlicz space $L_{M}$ for the modular in [5]. Duan and Chen gave a necessary condition (which is not sufficient) and a sufficient condition (which is not necessary) for prediction operators in $L_{M}$ in [11]. In 1995, Wang et al. [12] obtained a necessary and sufficient condition for an operator in $L_{M}$ to be a prediction operator. The purpose of this paper is to generalize the main result in [12] to Banach ideal spaces.

Theorem. Let $X(\Sigma)$ be a reflexive, strictly convex, and smooth Banach ideal space. Then the operator $T: X(\Sigma) \rightarrow X(\Sigma)$ is a prediction operator (i.e., there exists $\sigma$-sublattice $\Sigma^{\prime} \subset \Sigma$ such that $T(\cdot)=\pi\left(\cdot \mid X\left(\Sigma^{\prime}\right)\right)$ if and only if it satisf ies the following conditions:
(i) $T(T x)=T x(x \in X(\Sigma))$;
(ii) $r \in T X$ if $r \in R$;
(iii) $x, y \in T X$ and $\alpha, \beta \in R^{+}$imply $\alpha x+\beta y \in T X$;
(iv) $\left\{x_{n}\right\} \subset T X$ and $\left\|x_{n}-x\right\| \rightarrow 0(n \rightarrow \infty)$ imply $x \in T X$;
(v) $x \in T X$ and $r \in R$ imply $x \vee r, x \wedge r \in T X$;
(vi) $\chi_{A} \in T X$ and $\alpha, \beta \in R^{+}$and $r \in R$ imply $\|x-T x\| \leqslant \| x-\alpha T x-$ $\beta \chi_{A}+r \|$,
where $R$ is the set of all real numbers, $R^{+}$is the set of non-negative numbers, and $\chi_{A}$ is the characteristic function of the set $A$.

Proof. Necessity. Suppose $\Sigma^{\prime}$ is a $\sigma$-sublattice of $\Sigma$ and $T(\cdot)=$ $\pi\left(\cdot \mid X\left(\Sigma^{\prime}\right)\right)$. Obviously, $T X=X\left(\Sigma^{\prime}\right)$, so (i) is clear.

Note that $r \in R$ implies $r \in X\left(\Sigma^{\prime}\right)$ since $\phi, G \in \Sigma^{\prime}$, so (ii) is true.
It is easy to find that $X\left(\Sigma^{\prime}\right)$ is a closed convex cone in $X(\Sigma)$ since $\Sigma^{\prime}$ is a $\sigma$-sublattice. So (iii) and (iv) are true.

For brevity, we use the notation $\{x>a\}$ for the inverse image of the set $(a, \infty)$ under the mapping $x: G \rightarrow R$. Since

$$
\{x \vee r>a\}= \begin{cases}G & (a<r) \\ \{x>a\} & (a \geqslant r)\end{cases}
$$

and

$$
\{x \wedge r>a\}= \begin{cases}\{x>a\} & (a<r) \\ \phi & (a \geqslant r),\end{cases}
$$

we see that $x \vee r$ and $x \wedge r$ are $\Sigma^{\prime}$-measurable, and hence $x \vee r$, $x \wedge r \in X\left(\Sigma^{\prime}\right)=T X$. This establishes (v).

Since $T x \in T X, \chi_{A} \in T X, r \in T X$, and $T X$ is a convex cone, we have $\alpha T x+\beta \chi_{A}-r \in T X$. Therefore $\|x-T x\| \leqslant\left\|x-\left(\alpha T x+\beta \chi_{A}-r\right)\right\|$, and so (vi) holds.

Sufficiency. Let $\Sigma^{\prime}=\left\{A \in \Sigma: \chi_{A} \in T X\right\}$. Divide the proof into three steps as follows:

1. We prove that $\Sigma^{\prime}$ is a $\sigma$-sublattice of $\Sigma$.

By (ii) and (i), $\chi_{\Phi}=0=T 0=T \chi_{\Phi}, \chi_{G}=1=T 1=T \chi_{G}$. So $\chi_{\Phi}, \chi_{G} \in T X$, that is, $\Phi, G \in \Sigma^{\prime}$. If $A, B \in \Sigma^{\prime}$, then $\chi_{A}, \chi_{B} \in T X$. Observe that $\chi_{A \cup B}(t)=$ $\left(\chi_{A}(t)+\chi_{B}(t)\right) \wedge 1$ and $\chi_{A \cap B}(t)=\left(\chi_{A}(t)+\chi_{B}(t)-1\right) \vee 0$. Thus, by (iii) and (v), $\chi_{A \cup B}, \chi_{A \cap B} \in T X$. So $A \cup B, A \cap B \in \Sigma^{\prime}$.

Let $A=\bigcup_{n=1}^{\infty} A_{n}$, where $A_{n} \in \Sigma^{\prime}(n=1,2, \ldots)$ and $A_{1} \subset A_{2} \subset A_{3} \subset \cdots$. By Theorem 10 (Ogasawara) in [1, Chap. 10, Sect. 4] we find that $X(\Sigma)$ is order-continuous since $X(\Sigma)$ is reflexive. Therefore $\left\|\chi_{A}-\chi_{A_{n}}\right\| \rightarrow 0$ as $n \rightarrow \infty$ because $\mu\left(A / A_{n}\right) \rightarrow 0$ as $n \rightarrow \infty$. Using (iv), $\chi_{A} \in T X$ and it follows $A \in \Sigma^{\prime}$. If $A=\bigcap_{n=1}^{\infty} A_{n}$ where $A_{n} \in \Sigma^{\prime}(n=1,2, \ldots)$ and $A_{1} \supset A_{2} \supset A_{3} \supset \cdots$. By the above same arguments we can deduce that $A \in \Sigma^{\prime}$. Thus we have proved that $\Sigma^{\prime}$ is a $\sigma$-sublattice of $\Sigma$.
2. We prove that $X\left(\Sigma^{\prime}\right)=T X$.

Suppose $x \in T X$. For any $a \in R$ we write $y_{n}(t)=\left((3 / 2)^{n}(x(t)-a) \vee 0\right) \wedge 1$. By (iii) and (v) we see that $y_{n} \in T X(n=1,2, \ldots)$. Observe that $x(t) \leqslant a$ implies $y_{n}(t)=0=\chi_{\{x-a\}}(t)$ and $x(t)>a$ implies $y_{n}(t) \rightarrow 1=\chi_{\{x>a\}}(t)$ $(n \rightarrow \infty)$. We can deduce $\chi_{\{x>a\}}-y_{n} \downarrow 0(n \rightarrow \infty)$ and so $\left\|\chi_{\{x>a\}}-y_{n}\right\| \rightarrow 0$ $(n \rightarrow \infty)$ by the order-continuity of $X(\Sigma)$. It follows that $\chi_{\{x>a\}} \in T X$ from (iv). That is, $\{x>a\} \in \Sigma^{\prime}$ and then $x \in X\left(\Sigma^{\prime}\right)$. Hence $T X \subset X\left(\Sigma^{\prime}\right)$.

Conversely, suppose $x \in X\left(\Sigma^{\prime}\right)$. Then $\{x>a\} \in \Sigma^{\prime}$ and so $\chi_{\{x>a\}} \in T X$ for any $a \in R$.

Notice that $x \vee(-m) \downarrow x$ as $m \rightarrow \infty$ and thus $\|x \vee(-m)-x\| \rightarrow 0$ as $m \rightarrow \infty$. We need only prove $x \vee(-m) \in T X(m=1,2, \ldots)$ to show that $x \in T X$ by (iv). However, $x \vee(-m)=(x \vee(-m)+m)-m$ and $-m \in T X$, so
from (iii) we see that we only need to prove $x \vee(-m)+m \in T X$. Moreover, $x \vee(-m)+m \geqslant 0$ and $x \vee(-m)+m \in X\left(\Sigma^{\prime}\right)$, since $X\left(\Sigma^{\prime}\right)$ is a convex cone. In what follows, we need only prove that $0 \leqslant x \in X\left(\Sigma^{\prime}\right)$ implies $x \in T X$.

By Lemma 3 in [1, Chap. 4, Sect. 3],

$$
\left\|x-\sum_{k=0}^{n 2^{n}}\left(k / 2^{n}\right) \chi_{\left\{(k+1) / 2^{n} \geqslant x \geqslant k / 2^{n}\right\}}\right\| \rightarrow 0 \quad \text { as } \quad n \rightarrow \infty .
$$

However,

$$
\sum_{k=0}^{n 2^{n}}\left(k / 2^{n}\right) \chi_{\left\{(k+1) / 2^{n} \geqslant x \geqslant k / 2^{n}\right\}}=\left(1 / 2^{n}\right) \sum_{k=0}^{n 2^{n}} \chi_{\left\{x>k / 2^{n}\right\}} \in T X
$$

since $\chi_{\left\{x>k / 2^{n}\right\}} \in T X\left(n=1,2, \ldots ; k=1,2, \ldots, n 2^{n}\right)$ and by using (iii). Consequently $x \in T X$ by (iv). Hence $X\left(\Sigma^{\prime}\right) \subset T X$.
3. We prove that $T(\cdot)=\pi\left(\cdot \mid X\left(\Sigma^{\prime}\right)\right)$.

It is well known that $\pi\left(\cdot \mid X\left(\Sigma^{\prime}\right)\right)$ has meaning since $X(\Sigma)$ is reflexive and strictly convex.

If $x \in T X$, then $T x=x=\pi\left(x \mid X\left(\Sigma^{\prime}\right)\right)$.
Now suppose $x \notin T X$. Since $X(\Sigma)$ is smooth, $x-T x$ has a unique support functional $f_{x-T x} \in X(\Sigma)^{*}$ such that $\left\|f_{x-T x}\right\|_{*}=1$ and $f_{x-T x}(x-T x)$ $=\|x-T x\|$. Also $f_{x-T x}(T x)=\lim _{\lambda \rightarrow 0}(\|x-T x+\lambda T x\|-\|x-T x\|) / \lambda$ exists in the Gateaux sense. Nevertheless, $\|x-T x+\lambda T x\|=\|x-(1-\lambda) T x\|$ $\geqslant\|x-T x\|$ for $|\lambda|$ sufficiently small by (vi). This implies that $\lim _{\lambda \rightarrow 0^{+}}(\|x-T x+\lambda T x\|-\|x-T x\|) / \lambda \geqslant 0$ and $\lim _{\lambda \rightarrow 0^{-}}(\|x-T x+\lambda T x\|-$ $\|x-T x\|) / \lambda \leqslant 0$. Hence

$$
\begin{equation*}
f_{x-T x}(T x)=0 \tag{1}
\end{equation*}
$$

By the same arguments we can deduce that

$$
\begin{equation*}
f_{x-T x}(r)=0 \quad(r \in R) \tag{2}
\end{equation*}
$$

For any $\chi_{A} \in T X$, again by (vi) we have

$$
f_{x-T x}\left(-\chi_{A}\right)=\lim _{\lambda \rightarrow 0^{+}}(\|x-T x+\lambda T x\|-\|x-T x\|) / \lambda \geqslant 0
$$

and so $f_{x-T x}\left(\chi_{A}\right) \leqslant 0$.
For any $0 \leqslant u \in T X$, we have $\left\|u-\left(1 / 2^{n}\right) \sum_{k=0}^{n n^{n}} \chi_{\left\{u>k / 2^{n}\right\}}\right\| \rightarrow 0$ as $n \rightarrow \infty$. Since $\left\{u>k / 2^{n}\right\} \in \Sigma^{\prime}, \chi_{\left\{u>k / 2^{n}\right\}} \in T X$, we get $f_{x-T x}\left(\chi_{\left\{u>k / 2^{n}\right\}}\right) \leqslant 0 \quad(n=$ $\left.1,2, \ldots ; k=1,2, \ldots, n 2^{n}\right)$. Furthermore, $f_{x-T x}\left(\left(1 / 2^{n}\right) \sum_{k=0}^{n 2^{n}} \chi_{\left\{u>k / 2^{n}\right\}}\right) \leqslant 0$, and so $f_{x-T x}(u) \leqslant 0$.

In the general case, since $0 \leqslant u \vee(-m)+m \in T X(m=1,2, \ldots)$, we have $f_{x-T x}(u \vee(-m)+m) \leqslant 0(m=1,2, \ldots)$. Combining with Eq. (2), $f_{x-T_{x}}(u \vee(-m)) \leqslant 0(m=1,2, \ldots)$. But $\|u-(u \vee(-m))\| \rightarrow 0(m \rightarrow \infty)$. So

$$
\begin{equation*}
f_{x-T x}(u) \leqslant 0\left(u \in T X=X\left(\Sigma^{\prime}\right)\right) . \tag{3}
\end{equation*}
$$

By Eqs. (3) and (1), for any $u \in X\left(\Sigma^{\prime}\right)$ we have

$$
\begin{aligned}
\|x-T x\| & =f_{x-T x}(x-T x)=f_{x-T x}(x) \leqslant f_{x-T x}(x-u) \\
& \leqslant\left\|f_{x-T x}\right\|_{*}\|x-u\|=\|x-u\| .
\end{aligned}
$$

This means $T x=\pi\left(x \mid X\left(\Sigma^{\prime}\right)\right)$.

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