AN APPLICATION OF THE STRING AVERAGING METHOD TO ONE-SIDED BEST SIMULTANEOUS APPROXIMATION

HYANGJOO RHEE

ABSTRACT. For $t=1,2,\ldots,\ell$, let $I_t=(i_1^t,i_2^t,\ldots,i_{\ell(t)}^t)$ be an ordered $\ell(t)$ -tuple of numbers in $\{1, 2, \dots, \ell\}$ and let T_t be chosen from a finite composition of orthogonal projections

$$R_{i_1^t}, R_{i_2^t}, \dots, R_{i_{\ell(t)}^t}$$

 $R_{i_1^t}, R_{i_2^t}, \dots, R_{i_{\ell(t)}^t}$ acting on the normed linear space $C_1(X)$ to closed convex subset $S(f_{i_j^t})$ respectively. In this paper, we study the convergence of the sequence

$$x_i = \sum_{t=1}^{\ell} w_t T_t(x_{i-1}), \quad i = 1, 2, \ldots,$$

where $\sum_{i=1}^{\ell} w_i = 1$ and $w_i > 0$ for $i = 1, \dots, \ell$.

0. Introduction

This paper is concerned with one-sided best simultaneous approximation on

$$C(X) = \{f | f \text{ is a real valued continuous function on } X\},$$

where X is a compact subset of \mathbb{R}^N . We denote that $C_1(X) = (C(X), \|\cdot\|_1)$ with the $L_1(X,\mu)$ -norm where μ is a finite positive admissible measure defined on X. Then $C_1(X)$ is not a Banach space and it is a dense linear subspace of $L_1(X,\mu)$.

We define a norm on the space of all ℓ -tuples of functions in C(X) as follows:

For any ℓ elements f_1,\ldots,f_ℓ in C(X), let $F=(f_1,\ldots,f_\ell)$ and

$$||F|| = ||(f_1, \dots, f_\ell)|| = \max_{(w_1, \dots, w_\ell) \in A} || \sum_{i=1}^\ell w_i f_i ||_1$$

where $A = \{(w_1, \dots, w_\ell) \mid \sum_{i=1}^{\ell} w_i = 1 \text{ and } w_i > 0 \text{ for } i = 1, \dots, \ell\}.$

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Now suppose that an ℓ -tuple $F = (f_1, \ldots, f_{\ell})$ of functions in C(X) is given and S is a finitely generated subspace of C(X). We define the sets, for $i = 1, \ldots, \ell$,

$$S(f_i) := \{ s \in S \mid s(x) \le f_i(x) \text{ for } x \in X \}$$

and the set

$$S(F) := \bigcap_{i=1}^{\ell} S(f_i) := \bigcap_{i=1}^{\ell} \{ s \in S \mid s(x) \le f_i(x) \text{ for } x \in X \}$$

in normed linear space $C_1(X)$. By the definitions of $S(f_i)$ and S(F) above, it is trivial to show that S(F) is non-empty for every ℓ -tuple F of functions in C(X) if S contains a strictly positive function. Throughout this paper we shall restrict to those F for which S(F) is non-empty.

In this paper, we use a new algorithmic scheme for finding one-sided best approximation of an ℓ -tuple F from a set S(F). The new iterative algorithm (the string averaging method of Censor, Elfving & Herman [1] in approximation theory) has been used for solving the convex feasibility problem of finding a point x in the nonempty intersection $C = \bigcap_{i=1}^m C_i$ of finitely many closed and convex sets C_i in the Euclidean space \mathbb{R}^N .

1. One-sided best approximation

If $f^* \in S(F)$, we define

$$r(F, f^*) := \sup_{g \in F} ||f^* - g||.$$

The Chebyshev radius of F is defined by

$$rad_{S(F)}(F) := \inf_{f^* \in S(F)} r(F, f^*).$$

An element $f^* \in S(F)$ satisfying $r(F, f^*) = \operatorname{rad}_{S(F)}(F)$ is called a one-sided best simultaneous L_1 -approximation of F (cf. Park & Rhee [7]). If $\ell = 1$ (i. e., $F = \{f\}$ is singleton), it is called a one-sided best L_1 -approximation (cf. Pinkus [8] and Park & Rhee [7]).

Definition 1. Let V be a nonempty subset of a normed linear space Y and \mathbf{F} be a given family of closed and bounded subsets of Y.

A minimizing sequence $\{f_n\}$ of F in V is a sequence $\{f_n\}$ such that $r(F, f_n) \to \operatorname{rad}_V(F)$ has a subsequence which is convergent in V.

The set V is said to be *cent-compact relative to* \mathbf{F} if, for each $F \in \mathbf{F}$, each minimizing sequence $\{f_n\}$ in V.

We say that the set V satisfies the *center property* for \mathbf{F} if, for each $F \in \mathbf{F}$, there exists a one-sided best simultaneous L_1 -approximation of F.

Remark 1. For each ℓ -tuple $F = (f_1, \ldots, f_{\ell})$ of functions in $C_1(X)$, a minimizing sequence in S(F) is bounded, so the sequence has a subsequence which is convergent in S(F). Thus we can say that S(F) is cent-compact relative to F.

Mhaskar & Pai [6] has showed that if a closed set V in a normed space Y is cent-compact relative to Y, then V satisfies center property for each family of closed and bounded sets.

Consider a case when $\mathbf{F} = \{F\}$ is singleton and F is an ℓ -tuple of functions in $C_1(X)$. Then S(F) is center-compact relative to \mathbf{F} . Thus we have the following proposition.

Proposition 2. Let F is any ℓ -tuple of functions in of $C_1(X)$. Then S(F) satisfies center property for $F = \{F\}$.

2. An algorithm for orthogonal projections

Projection algorithmic schemes for the convex feasibility problem and for the best approximation problem are, in general, either sequential or simultaneous or block-iterative. In the following, we explain these terms in the framework of the algorithmic scheme proposed in this paper.

For each $t = 1, 2, ..., \ell$, let the *string* I_t be an ordered $\ell(t)$ -tuple of numbers in $\{1, 2, ..., \ell\}$ of the form

$$I_t = (i_1^t, i_2^t, \dots, i_{\ell(t)}^t).$$

We will assume that for any t, the components i_j^t $(j=1,2,\ldots,\ell(t))$ of I_t are distinct from each other and every element of $\{1,2,\ldots,\ell\}$ appears in at least one of the strings $I_t(t=1,2,\ldots,\ell)$.

Algorithmic Scheme. Let $w_t > 0$ (for $t = 1, 2, ..., \ell$), $\sum_{t=1}^{\ell} w_t = 1$ and $R_{i_j^t}$ be the metric projection onto $S(f_{i_i^t})$. We use iterative steps:

Given
$$F_0 = (f_1, ..., f_{\ell}),$$

(1) calculate, for all $t = 1, 2, \ldots, \ell$,

$$T_t f_t = R_{i_1^t} R_{i_2^t} \cdots R_{i_{\ell(t)}^t} f_t,$$

(2) and then calculate $F_1 = \sum_{t=1}^{\ell} w_t T_t f_t;$

given the current F_k $(k \ge 2)$, iterate

(3) calculate, for all $t = 1, 2, \ldots, \ell$,

$$T_t F_k = R_{i_1^t} R_{i_2^t} \cdots R_{i_{\ell(t)}^t} F_k,$$

(4) and then calculate $F_{k+1} = \sum_{t} w_t T_t F_k$.

We demonstrate our algorithmic scheme. For simplicity, we take $\ell = 1$. Then we get a one-sided L_1 -approximation (cf. Park & Rhee [7]).

In this framework, we get a simultaneous algorithm by the choice

$$I_t=(t), \quad t=1,2,\ldots,\ell,$$

and a sequential algorithm by the choice

$$I_t = (1, 2, \ldots, \ell).$$

For some case, the sequence $\{F_k\}$, generated by the algorithmic scheme, converges to a point $F^* \in S(F_0)$, which is a one-sided best simultaneous L_1 -approximation of F_0 in $S(F_0)$.

Now we turn to questions of when the sequence $\{F_k\}$ converges to some element in $S(F_0)$ and when the sequence converges to a one-sided best simultaneous L_1 -approximation to $F_0 = (f_1, \ldots, f_\ell)$.

The subspace S is called a one-sided simultaneous L_1 -unicity space if for each $F = (f_1, \ldots, f_\ell)$, there is a unique one-sided best simultaneous L_1 -approximation (cf. Park & Rhee [7]).

The next example satisfies that the sequence $\{F_k\}$ converges to a one-sided best simultaneous approximation of $F = (f_1, f_2)$.

Example 3. Let $F_0 = (\sin x, \cos x), X = [0, \pi]$ and $S = \mathbb{R}$ (i. e., S consists of all constant functions). Then S is a one-sided simultaneous L_1 -unicity space for $C_1(X)$ and

$$S(\sin x) = (-\infty, 0], S(\cos x) = (-\infty, -1].$$

So
$$S(F_0) = (-\infty, -1]$$
.

Let $I_t = (t)$; w_1 , w_2 are positive and $w_1 + w_2 = 1$. By Algorithm Scheme, we have

$$F_1 = -w_2,$$

$$F_2 = w_1(F_1) - w_2 = w_1(-w_2) - w_2 = -w_2(1+w_1),$$

$$F_3 = w_1(F_2) - w_2 = w_1(-w_2(1+w_1)) - w_2 = -w_2(1+w_1+w_1^2)$$

and therefore we have

$$F_n = w_1(F_{n-1}) - w_2 = -w_2(1 + w_1 + w_1^2 + \dots + w_1^{n-1})$$

for all $n \geq 1$. Then the sequence F_n converges to $F^* = -1 \in S(F_0)$ and F^* is a one-sided best simultaneous approximation of F_0 from $S(F_0)$.

3. Proofs of convergence

We now turn to prove our results concerning the convergence of $\{F_k\}$, generated by algorithmic scheme.

The framework in the previous section does not introduce relaxation parameters into the algorithm. However, according to Censor & Reich [3], we will do so for the special case of orthogonal projections.

We define, for $i = 1, 2, ..., \ell$, the algorithmic operators

$$R_i x = x + \alpha_i (P_{S(f_i)} x - x)$$
 for $x \in \mathbb{R}^{\ell}$

where α_i (0 < α_i < 2) are fixed for each set $S(f_i)$. The algorithmic operators R_i are called relaxed orthogonal projections.

Definition 2. An operator T is said to be *strictly nonexpansive*, if

$$||T(x) - T(y)|| < ||x - y|| \text{ or } T(x) - T(y) = x - y$$

for all x and y.

Definition 3. A continuous operator T is said to be *paracontracting*, if for any x and any fixed point y (i. e., y = T(y)),

$$||T(x) - y|| < ||x - y|| \text{ or } T(x) = x.$$

Remark 4. Obviously, a strictly nonexpansive operator is paracontracting. But the inverse implication does not hold (cf. Elsner, Koltracht & Neumann [4, Example 1]).

Next theorem shows that if ℓ -tuple $F_0 = (f_1, \ldots, f_{\ell})$ consist of constant functions then the sequence $\{F_k\}$ converges to some element in $S(F_0)$.

Theorem 5 (Censor, Elfving & Herman [1]). Let F_0 be an ℓ -tuple of elements in \mathbb{R} and S be a closed convex set in \mathbb{R}^{ℓ} . Then the sequence $\{F_k\}$, generated by the algorithmic scheme for the relaxed orthogonal projection, converges to a point $F^* \in S(F_0)$.

Corollary 6. Let F_0 be an ℓ -tuple of elements in \mathbb{R} and $S = \mathbb{R}$. Then the sequence $\{F_k\}$, in the above theorem, converges to a point $F^* \in S(F_0)$.

By the definition, F^* is a one-sided best simultaneous approximation for $F_0 = (f_1, \ldots, f_\ell)$ if and only if F^* satisfies that $F^* \in S(F_0)$ attains the supremum in $\sup_{f \in S(F_0)} \int_X f \, d\mu$. Obviously, if $\sup_{f \in S(F_0)} \int_X f \, d\mu = 0$ for all $f \in S(F_0)$, then S is not a one-sided simultaneous L_1 -unicity space for $C_1(X)$. Equivalently, if

$$\dim S = 1$$
, $\sup_{f \in S(F_0)} \int_X f \, d\mu \neq 0$ for some $f \in S$

and $S(F_0) \neq \emptyset$ for any $F_0 = (f_1, \ldots, f_\ell)$, then S is a one-sided simultaneous L_1 -unicity space for $C_1(X)$.

Proposition 7. Suppose that S is the set of all constant functions and let $\ell \geq 2$ and $I_t = (t)$. Then for any ℓ -tuple $F_0 = (f_1, \ldots, f_{\ell})$ in $C_1(X)$, the sequence $\{F_k\}$, constructed as in the algorithmic scheme with respect to relaxed projection operators, converges to a one-sided best simultaneous approximation of F_0 from $S(F_0)$.

Proof. Since S is a one-sided simultaneous L_1 -unicity space and $S(F_0) = (-\infty, \alpha)$ where

$$\alpha = \min_{i \in \{1,2,\dots,\ell\}} \{ \min_{x \in X} (f_i(x)) \},$$

we have

$$F_1 = \sum_{i=1}^{\ell} w_i \cdot \min_{x \in X} (f_i(x)).$$

If $F_1 \in S(F_0)$, then the proof is complete. If $F_1 \notin S(F_0)$, (i. e., there exists i_0 such that $F_1 \notin S(f_{i_0})$), then $F_2 \leq F_1$. Inductively, if $F_k \notin S(F_0)$ then $F_{k+1} \leq F_k$. So F_k is decreasing and convergent to a element F^* in $S(F_0)$ and F^* attains the supremum in $\int_X f \, d\mu$ for all $f \in S(F_0)$. Hence, F^* is a one-sided best simultaneous approximation of F_0 from $S(F_0)$.

4. Continuity of the projection $P_{S(\cdot)}(\cdot)$

For each ℓ -tuples F of C(X), S(F) is a closed and convex subset of $C_1(X)$. The metric projection $P: F \to P_{S(F)}(F)$ is a set valued map the approximating set depends on some ℓ -tuples F.

Now we consider a continuity of the projection $P_{S(\cdot)}(\cdot)$. By Remark 1, S(F) is a cent-compact and convex subset of S. Thus we can show that if the map $F \to S(F)$ is Hausdorff continuous at F_0 , then the projection $P_{S(\cdot)}(\cdot)$ is upper semicontinuous at F_0 .

Mabizela [5] proved that for each F, S(F) is a m-dimensional subspace of $C_1(X)$ with basis $\{e_1^F, \ldots, e_m^F\}$ with some conditions. Additionally if F_n converges to F_0 with $\|e_i^{F_n} - e_i^{F_0}\| \to 0$, then the map $S(\cdot)$ is Hausdorff continuous on F_0 .

Theorem 8. The metric projection $P_{S(\cdot)}(\cdot)$ is upper semicontinuous.

Proof. It suffices to show that the set

$$P_{S(\cdot)}^{-1}(E) := \{F \mid P_{S(F)}(F) \cap E \neq \emptyset\}$$

is closed in $C_1(X)$ for any closed set E. Let $\{F_n\} \subset P_{S(F)}^{-1}(E)$ be a sequence such that $\{F_n\}$ converge to F_0 with respect to Hausdorff metric, denote that $H(F_n, F_0) \to 0$ as $n \to \infty$. We can pick $u_n \in P_{S(F_n)}(F_n) \cap E$ for all $n \in \mathbb{N}$. Then $r(F_n, u_n) = \operatorname{rad}_{S(F_n)}(F_n)$ and

$$|r(F_0, u_n) - \operatorname{rad}_{S(F_0)}(F_0)| \le |r(F_0, u_n) - r(F_n, u_n)| + |\operatorname{rad}_{S(F_n)}(F_n) - \operatorname{rad}_{S(F_0)}(F_0)|$$

$$< 2H(F_n, F_0).$$

Thus, $r(F_0, u_n)$ approach to $\operatorname{rad}_{S(F_0)}(F_0)$. Hence $\{u_n\}$ is bounded and so there is a subsequence $\{u_k\}$ which is converge to $u_0 \in S$. Since the function $u \to r(F_0, u)$ is lower semicontinuous, we obtain

$$r(F_0, u_0) \le \liminf r(F_0, u_k) = \operatorname{rad}_{S(F_0)}(F_0).$$

Thus $u_0 \in P_{S(F_0)}(F_0) \cap E$, that is, $P_{S(\cdot)}^{-1}(E)$ is closed.

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Department of Liberal Arts, Duksung Women's University, 419 Ssangmun-dong, Dobong-gu, Seoul 132-714, Korea

 $Email\ address: \ {\tt rhj@center.duksung.ac.kr}$