

Continuity of the Best Approximation Operator for Restricted Range Approximations

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1. INTRODUCTION

In this paper we examine the behavior of the best uniform rational approximation operator in certain generalized weight function approximation problems. An introduction to this subject is given in [2].

Let X be a compact topological space, and for $f \in C(X)$ let

$$\|f\| = \max_{x \in X} |f(x)|.$$

Let P and Q be two finite dimensional linear subspaces of $C(X)$. In generalized rational approximation one is interested in approximating an $f \in C(X)$ by a function of the form $r = p/q$ where $p \in P, q \in Q$ and $q > 0$ on X .

A generalized weight function $W(x, y)$ is defined for $x \in X, y$ real, and has values in the extended reals. Specific examples and a number of results concerning generalized weight functions are given in ([1], [2], [3], [4]). In this paper we are concerned with the problem of finding a generalized rational function r which minimizes

$$\sup_{x \in X} |W[x, f(x) - r(x)]|. \tag{1}$$

The sections which follow give a number of results concerning (1), assuming various hypotheses on $W(x, y)$ and on the space of functions $P + rQ$ where r is a solution to the approximation problem. Here $P + rQ = \{p + rq : p \in P, q \in Q\}$.

Certain notations are used throughout the paper. Suppose for a fixed rational function r that $P + rQ$ has a basis g_1, \dots, g_n . Then for $x \in X$ we define a vector \hat{x} by

$$\hat{x} \equiv (g_1(x), g_2(x), \dots, g_n(x)). \tag{2}$$

The symbol 0 denotes the origin of Euclidean n -space. Suppose Y is a subset of X , and g is a real valued function defined on Y . Then

$$H\{g(y) \hat{y} : y \in Y\}$$

denotes the convex hull of the set of vectors $g(y) \hat{y}$ with $y \in Y$.

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If G is a linear subspace of $C(X)$, of dimension k , then G is called a Haar subspace iff every nonzero element of G has at most $k - 1$ zeros.

2. RESTRICTED RANGE APPROXIMATIONS

Let l and u be two elements of $C(X)$ satisfying

$$l(x) < u(x) \quad \forall x \in X.$$

Let $f^* \in C(X)$ be the function to be approximated, and define

$$R = \{r \equiv p/q : p \in P, q \in Q, q > 0, l \leq f^* - r \leq u\}. \tag{3}$$

In the discussion which follows we always assume that R is nonempty.

We shall consider a generalized weight function $W(x, y)$ with the following properties:

If $D = \{(x, y) : x \in X, y \text{ real}, l(x) \leq y \leq u(x)\}$ then:

- (a) $W(x, y)$ is continuous over D ;
 - (b) $\partial W(x, y)/\partial y$ is continuous over D and positive at each point (x, y) of D with $y \neq 0$;
 - (c) $(x, y) \in D \Rightarrow \text{sgn } W(x, y) = \text{sgn } y$;
 - (d) $x \in X$ and $y > u(x) \Rightarrow W(x, y) = \infty$;
 - (e) $x \in X$ and $y < l(x) \Rightarrow W(x, y) = -\infty$.
- } (4)

These hypotheses are satisfied, for example, in the problem considered in [4].

For notational convenience we write

$$E(f^* - r)(x) \equiv W[x, f^*(x) - r(x)].$$

We call $E(f^* - r)$ the weighted error function. Thus the problem (1) is to minimize

$$\sup_x |E(f^* - r)(x)| \equiv \|E(f^* - r)\|.$$

In restricted range approximations there are two types of critical points. For a particular $r \in R$ under consideration define:

$$\begin{aligned} X_{+1} &= \{x \in X : E(f^* - r)(x) = \|E(f^* - r)\|\} \\ X_{-1} &= \{x \in X : E(f^* - r)(x) = -\|E(f^* - r)\|\} \\ X_{+2} &= \{x \in X : E(f^* - r)(x) = u(x)\} \\ X_{-2} &= \{x \in X : E(f^* - r)(x) = l(x)\} \\ X_r &= X_{+1} \cup X_{-1} \cup X_{+2} \cup X_{-2}. \end{aligned}$$

In [1] it was shown that the cases $X_{+1} \cap X_{-2} \neq \emptyset$ and $X_{-1} \cap X_{+2} \neq \emptyset$ are exceptional, and not of general interest. Here we shall assume

$$X_{+1} \cap X_{-2} = X_{-1} \cap X_{+2} = \emptyset.$$

Then if $f^* \notin R$ we can define an integer valued function σ_r on X_r as follows

$$\sigma_r(x) = \begin{cases} \operatorname{sgn} E(f^* - r)(x) & x \in X_{+1} \cup X_{-1} \\ +1 & x \in X_{+2} \\ -1 & x \in X_{-2}. \end{cases}$$

For the remainder of this section we assume $f^* \notin R$. The following characterization theorem and lemma, which we shall need later, are established in [1].

THEOREM 1. *If $P + rQ$ is a Haar subspace then r is a best approximation to f^* iff*

$$0 \in H\{\sigma_r(x) \hat{x} : x \in X_r\}.$$

LEMMA 1. *If $P + rQ$ is a Haar subspace then*

$$0 \in H\{\sigma_r(x) \hat{x} : x \in X_r\}$$

iff there is no nonzero $h \in P + rQ$ such that $(\sigma_r, h)(x) \geq 0$ for all $x \in X_r$.

If r^* is a best approximation to f^* from R and $P + r^*Q$ is a Haar subspace, then r^* is unique [1]. In this situation we shall denote r^* by τf^* . We shall establish the continuity of the operator τ at a normal point $f^* \in C(X)$.

DEFINITION. $f^* \in C(X)$ is a normal point iff it has a best approximation r^* from R such that $P + r^*Q$ is a Haar subspace whose dimension = dimension $P + \text{dimension } Q - 1$.

Results concerning normal points can be found in ([5], [6], [7]). The first result we shall prove here is a *strong uniqueness theorem*.

THEOREM 2. *Let r^* be a best approximation to f^* from R . If f^* is normal then there exists an $\alpha > 0$ such that for all $r \in R$*

$$\|E(f^* - r)\| \geq \|E(f^* - r^*)\| + \alpha \|E(f^* - r^*) - E(f^* - r)\|. \quad (5)$$

Proof. (Note that this result is trivially true if $f^* \in R$.) We assume $f^* \notin R$ and that there is no α as stated. Then there exist sequences $\{r_n\} \subset R$ and $\{\alpha_n\}$, where $\alpha_n \rightarrow 0$ and

$$\alpha_n \|E(f^* - r^*) - E(f^* - r_n)\| = \|E(f^* - r_n)\| - \|E(f^* - r^*)\|.$$

Here $r_n = p_n/q_n$, $q_n > 0$, $\|p_n\| + \|q_n\| = 1$, and $r_n \neq r^*$. Since $l \leq f^* - r_n \leq u$, $\{r_n\}$ is

bounded. Here there is no loss of generality in assuming that there exist $p \in P, q \in Q$ such that $\|p\| + \|q\| = 1$ and $p_n \rightarrow p, q_n \rightarrow q$. We also can assume $r^* = p^*/q^*$ where $\|p^*\| + \|q^*\| = 1$. For simplicity of notation we shall write $\sigma(x) \equiv \sigma_{r^*}(x)$.

If $x \in X_{+1} \cup X_{-1}$ then

$$\begin{aligned} \alpha_n \|E(f^* - r^*) - E(f^* - r_n)\| &= \|E(f^* - r_n)\| - \|E(f^* - r^*)\| \\ &\geq \sigma(x) \{W[x, f^*(x) - r_n(x)] - W[x, f^*(x) - r^*(x)]\} \\ &= \sigma(x) \frac{\partial W[x, y_n(x)]}{\partial y} [r^*(x) - r_n(x)]. \end{aligned} \tag{6}$$

Here $y_n(x)$ is between $f^*(x) - r_n(x)$ and $f^*(x) - r^*(x)$. For the fixed x under consideration it might happen that zero is a point of accumulation of $\{f^*(x) - r_n(x)\}$. If that happens then by choosing subsequences one can assume $f^*(x) - r_n(x) \rightarrow 0$. Then for sufficiently large n ,

$$\sigma(x) [r^*(x) - r_n(x)] = \sigma(x) [r^*(x) - f^*(x) + f^*(x) - r_n(x)] \leq 0. \tag{7}$$

This uses the fact that

$$\sigma(x) [f^*(x) - r^*(x)] = \|(f^* - r^*)\| > 0.$$

Now by multiplying each side of (7) by $q_n(x)$ and taking limits, one concludes

$$0 \geq \sigma(x) [r^*(x)q(x) - p(x)]. \tag{8}$$

If $\{f^*(x) - r_n(x)\}$ does not have zero as a point of accumulation then there exists an N such that

$$d(x) \equiv \inf_{n \geq N} \frac{\partial W[x, y_n(x)]}{\partial y} > 0.$$

Hence for sufficiently large n it follows from (6) that

$$\frac{\alpha_n}{d(x)} \|E(f^* - r^*) - E(f^* - r_n)\| \geq \sigma(x) [r^*(x) - r_n(x)]. \tag{9}$$

Then by multiplying by $q_n(x)$ and taking limits one again obtains the inequality (8). That is, (8) holds for all $x \in X_{+1} \cup X_{-1}$.

For $x \in X_{+2} \cup X_{-2}$,

$$\sigma(x) [f^*(x) - r^*(x)] \geq \sigma(x) [f^*(x) - r_n(x)].$$

Hence

$$\sigma(x) [-r^*(x)q_n(x) + p_n(x)] \geq 0. \tag{10}$$

Taking limits we again conclude that (8) holds.

Since (8) holds for all $x \in X_r$ we obtain, using Lemma 1, $-r^*q + p \equiv 0$.

It then follows from ([5], p. 165) that $p^* \equiv p$, $q^* \equiv q$, and hence $r_n \rightarrow r^*$. We conclude that zero is not an accumulation point of $\{f(x) - r_n(x)\}$ when $x \in X_{+1} \cup X_{-1}$. Thus, since in any event $r_n \rightarrow r^*$ uniformly, there is no loss of generality in assuming there exists a $d > 0$ such that for all n and all $x \in X_{+1} \cup X_{-1}$,

$$d \leq \frac{\partial W[x, y_n(x)]}{\partial y}.$$

Since $q_n \rightarrow q^*$ uniformly, there exists a $\delta > 0$ such that for all n and all $x \in X$, $q_n(x) \geq \delta$. By a straightforward argument, using Lemma 1 and (10), it follows that there exists a $c > 0$ such that for all n ,

$$c \leq \max_{x \in X_{+1} \cup X_{-1}} \frac{\sigma(x)[r^*(x)q_n(x) - p_n(x)]}{\|r^*q_n - p_n\|}.$$

Using the above results in (6), we conclude

$$\begin{aligned} \alpha_n \|E(f^* - r^*) - E(f^* - r_n)\| &\geq dc \|r^*q_n - p_n\| \\ &\geq dc\delta \|r^* - r_n\|. \end{aligned}$$

An application of the mean value theorem to this inequality gives the existence of an $m > 0$ such that

$$m\alpha_n \|r_n - r^*\| \geq dc\delta \|r^* - r_n\|.$$

Since $r_n \neq r^*$ and $\alpha_n \rightarrow 0$, this yields the desired contradiction and completes the proof.

We now focus our attention on the continuity of τ at a normal point f^* . Let

$$F = \{f \in C(X) : l \leq f - \tau f^* \leq u\}. \tag{11}$$

For each $f \in F$, we consider the question of finding a solution to the problem of minimizing $\|E(f - r)\|$ for $r \in R$.

THEOREM 3. *Let f^* be a normal point of $C(X)$. Then there exists an $\alpha > 0$ such that $f_0 \in F$ and $\|f^* - f_0\| < \alpha$ imply that f_0 has at least one best approximation. Moreover, there exists a constant $\beta > 0$ such that for any best approximation r_0 to f_0 ,*

$$\|E(f^* - \tau f^*) - E(f_0 - r_0)\| \leq \beta \|f^* - f_0\|. \tag{12}$$

Proof. Let r^* be the best approximation to f^* . The search for a best approximation to f_0 may be confined to those $r_0 \in R$ for which

$$\|E(f_0 - r_0)\| \leq \|E(f_0 - r^*)\|.$$

Such r_0 satisfy (using the triangle inequality)

$$\begin{aligned} \|E(f^* - r^*) - E(f_0 - r_0)\| &\leq \|E(f^* - r^*) - E(f^* - r_0)\| \\ &\quad + \|E(f^* - r_0) - E(f_0 - r_0)\|. \end{aligned}$$

Using Theorem 2 and then the triangle inequality and other manipulations, it follows that the above is

$$\begin{aligned} &\leq \frac{1}{\alpha} [\|E(f^* - r_0)\| - \|E(f^* - r^*)\|] + \|E(f^* - r_0) - E(f_0 - r_0)\| \\ &\leq \frac{1}{\alpha} [\|E(f^* - r_0) - E(f_0 - r_0)\| + \|E(f_0 - r_0)\| - \|E(f^* - r^*)\|] \\ &\quad + \|E(f^* - r_0) - E(f_0 - r_0)\| \\ &\leq \frac{1}{\alpha} [\|E(f^* - r_0) - E(f_0 - r_0)\| + \|E(f_0 - r^*)\| - \|E(f^* - r^*)\|] \\ &\quad + \|E(f^* - r_0) - E(f_0 - r_0)\| \\ &\leq \frac{1}{\alpha} [\|E(f^* - r_0) - E(f_0 - r_0)\| + \|E(f_0 - r^*) - E(f^* - r^*)\|] \\ &\quad + \|E(f^* - r_0) - E(f_0 - r_0)\|. \end{aligned}$$

Application of the mean value theorem to each of the three “normed” quantities above, leads to the result (12). The proof is then completed by use of the methods in [5], p. 168, and [6].

It is worth noting that many generalized weight function approximations which do not have the restricted range condition can be considered to have it. For example, suppose $W(x, y)$ satisfies:

- (a) $\text{sgn } W(x, y) = \text{sgn } y$;
- (b) $W(x, y)$ and $\partial W(x, y)/\partial y$ are continuous;
- (c) $\partial W(x, y)/\partial y > 0$ when $y \neq 0$, and $\lim_{|y| \rightarrow \infty} |W(x, y)| = \infty$.

This allows us to select $u(x)$ sufficiently large, and $l(x)$ sufficiently small, so that $X_{+2} = \emptyset$ and $X_{-2} = \emptyset$. Then the results of Theorems 2 and 3 hold. These results are, thus, important if one is considering the computational aspects of this problem.

Next we consider the case where $P + (\tau f^*)Q$ is a Haar subspace but f^* is not necessarily a normal point of $C(X)$.

THEOREM 4. *Let $\{f_n\} \subset F$ and $\{r_n\} \subset R$ be two sequences such that*

$$f_n \rightarrow f^*$$

and

$$\|E(f_n - r_n)\| \rightarrow \|E(f^* - r^*)\|.$$

Here $r^* = \tau f^*$. If r_n is written in the normalized form $r_n = p_n/q_n$, with $\|p_n\| + \|q_n\| = 1$, then the sequence $\{(p_n, q_n)\}$ converges to the subspace

$$M \equiv \{(p, q) : p \in P, q \in Q, -p + r^*q \equiv 0\};$$

that is,

$$\text{distance } [M, (p_n, q_n)] \rightarrow 0.$$

Proof. If $r^* \equiv f^*$ we find $E(f_n - r_n) \rightarrow 0$, and hence by the properties of the weight function, $f_n - r_n \rightarrow 0$. Thus *a fortiori* we obtain the desired result.

If $f^* \neq r^*$ and the result is false then there exist subsequences of $\{f_n\}$ and $\{r_n\}$ which we do not relabel satisfying

- (a) there exist an $\epsilon > 0$ such that $\text{distance } [M, (p_n, q_n)] \geq \epsilon$ for all n ;
- (b) $p_n \rightarrow p, q_n \rightarrow q$ where $\|p\| + \|q\| = 1$.

For $x \in X_{+1} \cup X_{-1}$,

$$\begin{aligned} & \|E(f_n - r_n)\| - \|E(f^* - r^*)\| \\ & \geq \sigma_{r^*}(x) [E(f_n - r_n)(x) - E(f^* - r^*)(x)]. \end{aligned}$$

Using the same techniques as were employed in the proof of Theorem 2, one can verify that

$$0 \geq \sigma_{r^*}(x) [r^*(x)q(x) - p(x)]. \tag{13}$$

Since inequality (13) also holds for $x \in X_{+2} \cup X_{-2}$, it follows by Lemma 1 that

$$r^*q - p \equiv 0.$$

This contradicts the assumption that

$$\text{distance } [M, (p_n, q_n)] \geq \epsilon \quad \text{for all } n$$

and completes the proof.

For the remainder of this section we specialize to the situation where $X = [a, b]$. We make the assumption that for each nonzero $q \in Q$, the set of zeros of q is of measure zero.

THEOREM 5. *If $\{r_n\} \subset R$ and $\{f_n\} \subset F$ are such that $r_n = p_n/q_n, \|p_n\| + \|q_n\| = 1, (p_n, q_n) \rightarrow M$, and $f_n \rightarrow f^*$, then $E(f_n - r_n) \rightarrow E(f^* - r^*)$ in measure. Here $M = \{(p, q) : p \in P, q \in Q, -p + r^*q \equiv 0\}$.*

Proof. Assume the contrary. We can then find subsequences of $\{r_n\}$ and $\{f_n\}$, which we do not relabel, such that

- (a) There exist an $\epsilon > 0$ and a positive integer k such that if

$$B_n \equiv \{x : |E(f_n - r_n)(x) - E(f^* - r^*)(x)| > 1/k\}$$
 then the measure of B_n is greater than ϵ for all n ;
- (b) $p_n \rightarrow p, q_n \rightarrow q$ where $\|p\| + \|q\| = 1$.

Since $\|p\| + \|q\| = 1$ and $-p + r^*q \equiv 0$, we conclude that $q \neq 0$. Let

$$X_0 = \{x: q(x) \neq 0\}.$$

By hypothesis the measure of X_0 is $b - a$. Choose a closed set $X_1 \subset X_0$ such that the measure of X_1 is $b - a$. On X_1 , $E(f_n - r_n) \rightarrow E(f^* - r^*)$ uniformly. Thus for large n , $B_n \cap X_1 = \emptyset$, which implies that B_n has measure zero. This is a contradiction.

The following result is then clear.

THEOREM 6. *If r^* is a best approximation to f^* and $P + r^*Q$ is a Haar subspace, then for every pair of sequences $\{r_n\} \subset R$ and $\{f_n\} \subset F$ such that $f_n \rightarrow f$ and $\|E(f_n - r_n)\| \rightarrow \|E(f^* - r^*)\|$, $E(f_n - r_n) \rightarrow E(f^* - r^*)$ in measure.*

3. RATIONAL APPROXIMATION WITH INTERPOLATION

We turn now to a different sort of restricted range approximation. Using the ordinary uniform norm as a measure of error we are interested in finding a best rational approximation which interpolates $f(x)$ on a prescribed point set. To be more specific, let $\{x_1, \dots, x_k\} \subset X$, where $k \leq \text{dimension } P$, be a given set of points. For $f \in C(X)$ let

$$R_f = \{r \equiv p/q: p \in P; q \in Q; q > 0; r(x_i) = f(x_i), \quad i = 1, \dots, k\}$$

Then we call $r^* \in R_f$ a best approximation to f from R_f iff

$$\text{distance}(R_f, f) = \|f - r^*\|.$$

For each $r \in R_f$ define

$$S_r = \{-p + rq: p \in P; q \in Q; (-p + rq)(x_i) = 0, \quad i = 1, \dots, k\}.$$

DEFINITION. S_r is called an interpolating Haar subspace iff every nonzero element in S_r has at most $d(r) - 1$ zeros distinct from $\{x_1, \dots, x_k\}$. $d(r)$ is the dimension of the subspace S_r .

Clearly if $P + rQ$ is a Haar subspace, then S_r is an interpolating Haar subspace. The following theorem and lemma are given in [8].

THEOREM 7. *r is a best approximation to f from R_f iff*

$$0 \in H\{\sigma(x) \hat{x}: x \in X_r\}$$

where

$$\sigma(x) = \text{sgn}[f(x) - r(x)], \quad X_r = \{x \in X: |f(x) - r(x)| = \|f - r\|\},$$

Here $\hat{x} \equiv (g_1(x), g_2(x), \dots, g_n(x))$, where g_1, g_2, \dots, g_n is a basis of S_r .

LEMMA 2. If r is a best approximation to f from R_f , where $r \neq f$ and S_r is an interpolating Haar subspace, then $h \in S_r$ and $\sigma(x)h(x) \geq 0$ for all $x \in X$, imply $h \equiv 0$.

In [8], under the assumption that the dimension of the interpolating Haar subspace S_r is (dimension $P + \text{dimension } Q - 1 - k$), the Lipschitz continuity of the best approximation operator at f was demonstrated. In general we will show that only convergence in measure can be expected.

THEOREM 8. Let r be a best approximation to f from R_f and assume S_r is an interpolating Haar subspace. Let $\{r_n\}$ and $\{f_n\}$ be two sequences with the properties:

- (a) $r_n \in R_{f_n}$, where $r_n = p_n/q_n$ and $\|p_n\| + \|q_n\| = 1$;
- (b) $f_n \rightarrow f$;
- (c) $\|f_n - r_n\| \rightarrow \|f - r\|$.

Define $M = \{(p, q) \in P \times Q : -p + rq \equiv 0\}$. Then

$$\text{distance} [(p_n, q_n), M] \rightarrow 0$$

Proof. For the case $r \equiv f$, the result is clear. If $r \neq f$, assume that the result is false. Then (by taking subsequences if necessary) there exists an $\epsilon > 0$ such that

$$\text{distance} ((p_n, q_n), M) \geq \epsilon \tag{14}$$

for all n . By taking further subsequences we can secure that $p_n \rightarrow p$ and $q_n \rightarrow q$. Now, for each interpolating point x_i ,

$$-p_n(x_i) + q_n(x_i) f_n(x_i) = 0.$$

Since $f_n(x_i) = r_n(x_i)$, one finds by taking the limit,

$$-p(x_i) + q(x_i) r(x_i) = 0.$$

Hence $-p + rq \in S_r$. By the same argument used in Theorem 2,

$$-p(x) + q(x) r(x) = 0$$

for each $x \in X_r$. Hence by Lemma 2

$$-p + rq \equiv 0.$$

This contradicts (14).

THEOREM 9. If $r^* \in R_f$, and $P + r^*Q$ is a Haar subspace, then there exists a $\gamma > 0$ such that $\|f - g\| < \gamma$ implies that R_g is nonempty. Furthermore, if $f_n \rightarrow f$ and $\|f - f_n\| < \gamma$, there exist $r_n \in R_{f_n}$ such that $r_n \rightarrow r^*$.

Proof. Consider the system of equations for p and q

$$-p(x_i) + g(x_i)q(x_i) = 0 \quad i = 1, \dots, k.$$

By hypothesis, this system can be solved in a neighborhood of $p = p^*$, $q = q^*$ and $g = f$ for a p and q such that if $r = p/q$, $r \in R_g$ and r is close to r^* .

COROLLARY. *Under the same hypotheses as in the previous theorem, $f_n \rightarrow f$ implies distance $(R_{f_n}, f_n) \rightarrow$ distance (R_f, f) .*

Now if we specialize to the case where $X = [a, b]$ and assume for each non-zero $q \in Q$, that the set of zeros of q has measure zero, we find, pursuing the same ideas as in the restricted range case, that:

THEOREM 10. *Assume r is a best approximation to f from R_f and S_r is an interpolating Haar subspace. Then if $\{r_n\}$ and $\{f_n\}$ are two sequences such that $f_n \rightarrow f$, $r_n \in R_{f_n}$ and $\|f_n - r_n\| \rightarrow \|f - r\|$, then $r_n \rightarrow r$ in measure.*

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