

## Varying Norms and Constraints

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The effects of perturbing norm and constraints in linear approximation are considered. © 1987 Academic Press, Inc.

Let  $\phi_1, \dots, \phi_n$  be linearly independent and define the linear approximating function

$$L(A) = \sum_{i=1}^n a_i \phi_i.$$

Let  $\|\cdot\|_k$  be a seminorm on the space  $C$  and let  $K_k$  be a nonempty closed subset of  $n$ -space. The  $k$ th problem of constrained approximation is, given  $f \in C$ , to find a parameter  $A^k$  to minimize  $e_k(A) = \|f - L(A)\|_k$  subject to the constraint  $A \in K_k$ .

We generalize a perturbation result of Kripke [3] for approximation with no constraint (i.e.,  $K_k = n$ -space). We use more elementary arguments.

**THEOREM.** *Let  $\|\cdot\|_0$  be a norm on  $C$  and for each  $g \in C$ , let  $\|g\|_k \rightarrow \|g\|_0$ . Let  $A^k \in K_k$  and  $A^k \rightarrow A^0$  imply  $A^0 \in K_0$ . Let every element of  $K_0$  be a limit point of a sequence  $\{A^k\}$ ,  $A^k \in K_k$ . Let  $A^k$  be best in the  $k$ th problem of constrained approximation. Then  $\{A^k\}$  has an accumulation point and any accumulation point  $A^0$  is best with respect to  $\|\cdot\|_0, K_0$ .*

*Proof.* Define

$$\|A\|_c = \max\{|a_i|: 1 \leq i \leq n\}.$$

Suppose  $\|A^k\|_c$  is unbounded. By taking a subsequence if necessary we can assume  $\|A^k\|_c \geq k$ . Define  $B^k = A^k / \|A^k\|_c$ ; then  $\|B^k\|_c = 1$ . By taking a subsequence if necessary, we can assume  $B^k \rightarrow B^0$ . Now

$$\begin{aligned} \|f - L(A^k)\|_k &\geq \|L(A^k)\|_k - \|f\|_k = \|A^k\|_c \|L(B^k)\|_k - \|f\|_k \\ &\geq k \|L(B^k)\|_k - \|f\|_k. \end{aligned} \tag{1}$$

Now

$$\|L(B^k)\|_k \geq \|L(B^0)\|_k - \|L(B^k - B^0)\|_k, \quad (2)$$

$$\|L(B^k - B^0)\|_k \leq \|B^k - B^0\|_c \sum_{j=1}^n \|\phi_j\|_k. \quad (3)$$

For all  $k$  sufficiently large  $\|\phi_j\|_k \leq 2\|\phi_j\|_0$ ; hence by (3),

$$\|L(B^k - B^0)\|_k \rightarrow 0;$$

hence by (2)

$$\liminf_{k \rightarrow \infty} \|L(B^k)\|_k \geq \|L(B^0)\|_0.$$

By (1) we have

$$\|f - L(A^k)\|_k \rightarrow \infty. \quad (4)$$

But for any coefficient  $D^0$  and  $\{D^k\} \rightarrow D^0$ ,  $D^k \in K_k$ ,

$$\limsup_{k \rightarrow \infty} \|f - L(D^k)\|_k \leq \|f - L(D^0)\|_0. \quad (5)$$

To prove this,

$$\begin{aligned} \|f - L(D^k)\|_k &\leq \|f - L(D^0)\|_k + \|L(D^0 - D^k)\|_k \\ &\leq \|f - L(D^0)\|_k + \sum_{i=1}^n |d_i^0 - d_i^k| \|\phi_i\|_k. \end{aligned}$$

Now (5) contradicts (4), so  $\|A^k\|_c$  is bounded. Hence  $A^k$  has an accumulation point  $A^0$ . By taking a subsequence if necessary we can assume  $\{A^k\} \rightarrow A^0$ . Our hypotheses on  $K_k$  and  $K_0$  ensure that  $A^0 \in K_0$ . Finally, suppose  $A^0$  is not best with respect to  $\|\cdot\|_0$  and  $K_0$ . Then there is  $B \in K_0$  and  $\varepsilon > 0$  with

$$\|f - L(B)\|_0 < \|f - L(A^0)\|_0 - \varepsilon,$$

which implies for all  $k$  sufficiently large and  $\{A^k\} \rightarrow A^0$ ,  $\{B^k\} \rightarrow B$ ,

$$\begin{aligned} B^k &\in K_k, \\ \|f - L(B^k)\|_k &< \|f - L(A^k)\|_k - \varepsilon/2. \end{aligned} \quad (6)$$

To derive (6),

$$\begin{aligned}
 & \|f - L(B^k)\|_k - \|f - L(A^k)\|_k \\
 & \leq \|f - L(B)\|_k + \|L(B - B^k)\|_k - \|f - L(A^0)\|_k + \|L(A^k - A^0)\|_k \\
 & \leq \|f - L(B)\|_k - \|f - L(A^0)\|_k \\
 & \quad + \sum_{i=1}^n |b_i - b_i^k| \|\phi_i\|_k + \sum_{i=1}^n |a_i^0 - a_i^k| \|\phi_i\|_k.
 \end{aligned}$$

The first two terms tend to  $\|f - L(B)\|_0 - \|f - L(A^0)\|_0$  and the last two terms tend to 0. The proof is finished as (6) contradicts optimality of  $A^k$  in the  $k$ -problem.

*Remark.* With a little more work we can prove

$$\|f - L(A^k)\|_k \rightarrow \|f - L(A^0)\|_0. \quad (*)$$

Suppose not; then in view of (5) it suffices to prove

$$\liminf_{k \rightarrow \infty} \|f - L(A^k)\|_k \geq \|f - L(A^0)\|_0 \quad (7)$$

Suppose not; then we can assume

$$\|f - L(A^k)\|_k < \|f - L(A^0)\|_0 - \delta. \quad (8)$$

But

$$\begin{aligned}
 \|f - L(A^k)\|_k & \geq \|f - L(A^0)\|_k - \|L(A^k - A^0)\|_k \\
 & \geq \|f - L(A^0)\|_k - \sum_{i=1}^n |a_i^k - a_i^0| \|\phi_i\|_k.
 \end{aligned}$$

This contradicts (8), proving (7); (\*) may prove essential in computational work, as  $\|\cdot\|_0$  may be difficult to compute, e.g., discretization.

*Remark.* If  $f$  has a unique best approximation  $L(A^0)$  in the 0-problem,  $\|L(A^0) - L(A^k)\|_0 \rightarrow 0$ .

It is expected that in most mathematical applications of the theorem either the norm will be fixed or the constraints will be fixed, but both will be perturbed in computation.

To fully exploit the theorem we need results on nearby norms. Perhaps because Kripke's result is little known, such results have not appeared. Indeed, papers have appeared subsequently [5, 9, 10] with results that could be derived in one line from Kripke's result. Kripke cites discrete  $L_p$  norms,  $1 \leq p \leq \infty$ , as limits of  $L_{\rho(k)}$  norms. The same holds for  $1 \leq p < \infty$

and intervals providing functions are continuous. It is clear that if we let  $C$  be the continuous functions on an interval  $I$  containing intervals  $I_0, I_1, \dots, I_k, \dots$  then the limit of  $L_p$  norms on intervals  $\{I_k\} \rightarrow I_0$  is the  $L_p$  norm on interval  $I_0$  (compare with the author's [8]). Let  $C$  be the continuous functions in an interval  $I$  and  $w_0, w_1, \dots, w_k, \dots$  be positive continuous weight function on  $I$ ; then the limit of  $w_k$ -weighted  $L_p$  norms on  $I$  is the  $w_0$ -weighted  $L_p$  norm on  $I$  (with care this result might be applicable to some  $w_0$ 's with zeros). The above three observations might be combined. Let  $C$  be the continuous functions on interval  $I$  and  $\{Q_k\}$  a sequences of quadrature rules such that  $Q_k(g) \rightarrow \int g$  for  $g \in C$ : if we fix  $p$  in  $[1, \infty)$  and choose  $\|g\|_k = [Q_k(|g|^p)]^{1/p}$ , then  $\|g\|_k$  tends to the  $L_p$  norm of  $g$  on  $I$ . Kripke [3, p. 104] thought some such result held. For some sequences of quadrature rules it is only necessary that  $g$  be Riemann integrable. It is known that the limit of  $L_{p(k)}$  norms for  $g \in C[0, 1]$  is the  $L_\infty$  norm [12, Problem 4]. Extensions to Bacopoulos-type norms [11] and Moursund-type problems [13, 14] are straightforward.

The hypotheses of sentences 2 and 3 of the theorem are related to the hypotheses of [2]. It is shown in Appendix 2 that they are the most general possible. Levasseur has perturbation hypotheses in the appendix to his thesis [4]. They do not cover varying interpolatory constraints or varying restricted range, but ours do.

In some problems constraints do not change, that is,  $K_0 = K_1 = \dots = K_k = \dots$  and the hypotheses on the constraints are automatically satisfied. These problems include unchanging interpolatory constraints, for example, Lagrange-type or Hermite-type, and co-positive, co-monotone, co-convex constraints [1]. With varying interpolatory constraints matters may be less satisfactory.

EXAMPLE 1. Approximate  $f(x) = x^2$  by linear combinations of  $\{1, x\}$ . We require interpolation on  $\{0, 1/k\}$ : the only interpolant is  $k^3x$  and no limit exists as  $k \rightarrow \infty$ . But any multiple of  $x$  interpolates  $f$  at zero.

EXAMPLE 2. Approximate  $f(x) = x^2$  by multiples of  $x$ . We require interpolation on  $\{1/k\}$ : apply the rest of the above example.

In Example 1 we have coalescing of nodes and in Example 2 a non-Haar approximant.

We now establish the hypotheses of the main theorem for the case in which  $C$  is the continuous functions on compact  $X$ ,  $L$  satisfies the Haar uniqueness condition, and points of interpolation do not coalesce. Let  $\{x_i^k\} \rightarrow x_i^0$ ,  $i = 1, \dots, j$ , and

$$K_k = \{A: L(A)(x_i^k) = f(x_i^k), i = 1, \dots, j\}$$

where  $j$  is a fixed number between 1 and  $n$ . Let  $A^k \in K_k$  and  $\{A^k\} \rightarrow A^0$ . By uniform convergence of  $L(A^k)$  to  $L(A^0)$  we must have  $L(A^0)(x_i^0) = f(x_i^0)$ ,  $i = 1, \dots, j$ , establishing the hypotheses of the second sentence of the main theorem. Let  $Y$  be a set of  $n - j$  points distinct from  $\{x_i^0, \dots, x_j^0\}$ . Let  $A^0 \in K_0$  be given and choose  $A^k$  to satisfy

$$\begin{aligned} L(A^k)(x_i^k) &= f(x_i^k) & i = 1, \dots, j \\ L(A^k)(y) &= L(A^0, y) & y \in Y. \end{aligned}$$

Apply the theorem of the Appendix.

With a number of unchanging restricted range constraints [1, p. 62ff] on function or derivatives (a sample one is

$$\mu \leq L^{(r)}(A) \leq v),$$

$K_0 = \dots = K_k = \dots$  and the hypotheses on  $K_0, K_k$  are satisfied automatically.

Consider one changing restricted range constraint on function or derivative, namely,

$$\mu_k \leq L^{(r)}(A) \leq v_k.$$

Let  $\mu_k \rightarrow \mu_0$  uniformly and  $v_k \rightarrow v_0$  uniformly. Providing  $K_0$  is nonempty, the hypothesis of the second sentence of the theorem is satisfied: suppose without loss of generality that  $L^{(r)}(A^0)(x) < \mu_0(x) - \varepsilon$ ; then for all  $k$  sufficiently large

$$L^{(r)}(A^k)(x) < \mu_k(x) - \varepsilon/2, \tag{9}$$

a contradiction. It does not appear possible to verify the hypothesis of the third sentence of the theorem without further assumptions: we might assume  $\mu_k \leq \mu_0$ ,  $v_0 \leq v_k$ , which makes it automatic. Combining several changing restricted range constraints on functions and derivatives is straightforward.

The case of a changing restricted range *on one side* can be handled without very restrictive assumptions. Let  $C$  be the continuous functions on  $X$ , a compact subset of the real line and

$$K_k = \{A: L^{(r)}(A) \geq \mu_k\}, \tag{10}$$

and  $\mu_k \rightarrow \mu_0$  uniformly on  $X$ .

LEMMA. *Let there exist  $B$  such that  $L^{(r)}(B) > 0$  on  $X$ . Then any  $A \in K_0$  is a limit point of a sequence  $\{A^k\}$ ,  $A^k \in K_k$ .*

*Proof.* For conciseness we suppress the superscript ( $r$ ) on  $L$ . Suppose not then there is  $\delta > 0$  such that for all  $A^k \in K_k$

$$\|A - A^k\|_c > \delta \quad (11)$$

for all  $k$  sufficiently large. Now let  $\lambda_k$  be the smallest number  $\lambda \geq 0$  such that

$$L(A + \lambda B) \geq \mu_k. \quad (12)$$

$\lambda_k$  is well defined as  $L(A + \lambda B)$  becomes indefinitely large as  $\lambda \rightarrow \infty$  since  $L(B) > 0$  and the infimum of  $\lambda$ 's  $\geq 0$  satisfying (12) must satisfy (12). If  $\{\lambda_k\}$  had zero as an accumulation point, (11) would be contradicted so we will assume there is  $\varepsilon > 0$  such that  $\lambda_k \geq \varepsilon$  for all  $k$  sufficiently large. By taking a subsequence, we can assume  $\{\lambda_k\} \rightarrow \lambda_0 \geq \varepsilon$ . By choice of  $\lambda_k$ ,  $L(A + \lambda_k B)$  must graze  $\mu_k$ ; that is, there is  $x_k$  such that

$$L(A + \lambda_k B)(x_k) = \mu_k(x_k) \quad (13)$$

for if this did not happen, we could reduce  $\lambda_k$  and still satisfy (12). By compactness of  $X$ , the sequence  $\{x_k\}$  has an accumulation point  $x_0$ ; assume  $\{x_k\} \rightarrow x_0$ . As  $L(A + \lambda_k B) \rightarrow L(A + \lambda_0 B)$  uniformly on  $X$ ,  $L(A + \lambda_0 B)(x_0) = \mu_0(x_0)$  by (13). But as  $L(A) \geq \mu_0$  and  $L(B) > 0$  we have a contradiction, proving the lemma.

The lemma establishes the hypothesis of the third sentence of the theorem. The hypothesis of the second sentence is handled by preceding arguments.

The general case of changing restraints on two sides, if it can be handled at all, will require more subtle analysis.

EXAMPLE. Let  $C$  be the continuous functions on  $X = [0, 1]$ . Choose

$$\begin{aligned} v_k(x) &= 0 & 0 \leq x \leq 1/k \\ &= x - 1/k & x > 1/k \\ \mu_k &= -v_k \end{aligned}$$

By Dini's theorem  $v_k \rightarrow v_0 = x$  uniformly on  $X$  and  $\mu$  is similar. Now let us consider approximation by first-degree polynomials; then  $\{0\}$  is the only approximant in the restricted range of  $\mu_k, v_k$ . But the approximants in the restricted range of  $\mu_0, v_0$  are  $\{rx: |r| \leq 1\}$ .

This example suggests that no general theory is possible if restraining functions touch.

Even if restraining functions do not touch, a general theory may not be

possible without conditions on approximants. Take the above example, except  $\mu_0 = \mu_k = -1$  and approximants are multiples of  $x$ .  $\{0\}$  is the only nonnegative approximation restrained by  $\mu_k, \nu_k$  but  $\{rx: 0 \leq r \leq 1\}$  is restrained by  $\mu_0, \nu_0$ . If we completely neglect the lower restraint, we have an example for which the  $L^{(r)}(B)$  condition of the preceding lemma is necessary.

A condition on approximants which does guarantee that the hypotheses of the theorem are satisfied is the ASSUMPTION of the author's paper [2, 196].

### APPENDIX 1: VARYING INTERPOLATORY CONSTRAINTS

**THEOREM.** *Let  $L$  have the Haar property. Let  $L(A^k)$  take the values  $y_1^k, \dots, y_n^k$  at points  $x_1^k, \dots, x_n^k$  (all distinct), respectively. Let  $y_j^k \rightarrow y_j^0, j = 1, \dots, n$ , and  $x_j^k \rightarrow x_j^0, j = 1, \dots, n$ . Then  $A^k \rightarrow A^0$ .*

*Proof.* First we must show  $\{A^k\}$  is bounded. Suppose not; then by taking a subsequence if necessary we can assume  $\|A^k\|_c > k$ . By standard arguments, e.g., that of Rice [6, 24–25], this leads to a contradiction. Let  $B$  be any accumulation point of  $\{A^k\}$  not equal to  $A^0$ ; then by taking a subsequence if necessary we can assume  $\{A^k\} \rightarrow B$ . By uniform convergence of  $L(A^k)$  to  $L(B)$  we must have  $L(B)(x_j^0) = y_j^0, j = 1, \dots, n$ . But by the Haar assumption interpolation is unique.

*Remark.* Tornheim [7] proved this for  $X$  an interval and real values.

*Remark.* This could be generalized to Hermite-type interpolation by replacing the Haar condition.

### APPENDIX 2: NECESSITY OF HYPOTHESES

If we do not have the hypotheses of the second and third sentences of the main theorem satisfied, the conclusions of the theorem do not hold. Suppose (to contradict the second sentence)  $D^k \in K_k$  and  $\{D^k\} \rightarrow D^0$  not in  $K_0$ . Approximate  $f = L(D^0)$ ; then by (5)  $\{A^k\} D^0$  for  $A^k$  best in the  $k$ -problem. Suppose (to contradict the third sentence)  $D^0 \in K_0$  is not a limit point of any sequence  $\{D^k\}, D^k \in K_k$ . Approximate  $f = L(D^0)$ ; then  $A^k$  best in the  $k$ -problem implies that  $\|A^k - D^0\|_c$  is bounded away from zero for  $k$  large.

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