

## ON BEST SIMULTANEOUS APPROXIMATION IN NORMED LINEAR SPACES

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1. Let  $S$  be a non-empty family of real valued continuous functions on  $[a, b]$ . Diaz and McLaughlin [1], [2], and Dunham [3] have considered the problem of simultaneously approximating two continuous functions  $f_1$  and  $f_2$  by elements of  $S$ . If  $\|\cdot\|$  denotes the supremum norm, then the problem is to find an element  $s^* \in S$ , if it exists, for which

$$\max(\|f_1 - s^*\|, \|f_2 - s^*\|) = \inf_{s \in S} \max(\|f_1 - s\|, \|f_2 - s\|).$$

We shall study the above problem in general normed linear spaces.

DEFINITION 1.1. Let  $X$  be a normed linear space and  $K$  a subset of  $X$ . Given any two elements  $x_1, x_2 \in X$  define:

$$d(x_1, x_2; k) = \inf_{k \in K} \max(\|x_1 - k\|, \|x_2 - k\|).$$

An element  $k^* \in K$  is said to be a best simultaneous approximation to  $x_1$  and  $x_2$  if:

$$d(x_1, x_2; k) = \max(\|x_1 - k^*\|, \|x_2 - k^*\|)$$

2. First we show that the best simultaneous approximation exists if the set  $K$  is a finite dimensional subspace of the normed linear space  $X$ .

LEMMA 2.1. Let  $x_1, x_2 \in X$  and let  $k \in X$ . Then  $\phi(k) \equiv \max(\|x_1 - k\|, \|x_2 - k\|)$  is a continuous functional on  $X$ .

**Proof.** Since the norms  $\|x_1 - k\|, \|x_2 - k\|$  are continuous functionals of  $k$  on  $X$ ,  $\phi(k)$  is clearly a continuous functional.

LEMMA 2.2. If  $K$  is a finite dimensional subspace of a normed linear space  $X$ , then there exists a best simultaneous approximation  $k^* \in K$  to  $x_1, x_2 \in X$ .

**Proof.** Let  $\rho = \max(\|x_1\|, \|x_2\|)$ . Consider the spheres  $S(x_1, \rho), S(x_2, \rho)$  in  $K$  and write:

$$S = S(x_1, \rho) \cup S(x_2, \rho).$$

Then

$$\inf_{k \in S} \max(\|x_1 - k\|, \|x_2 - k\|) = \inf_{k \in K} \max(\|x_1 - k\|, \|x_2 - k\|) \leq \rho.$$

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Since  $S$  is compact, the continuous functional  $\phi(k)$  defined on  $S$  attains its minimum over  $S$ . If  $\min \phi(k) = \phi(k^*)$  then the element  $k^*$  is a best simultaneous approximation to  $x_1$  and  $x_2$ , and the lemma is proved.

**LEMMA 2.3.** *Let  $K$  be a convex subset of  $X$ , and  $x_1, x_2 \in X$ . If  $k_1, k_2 \in K$  are best simultaneous approximations to  $x_1, x_2$  by the elements of  $K$ , then:  $\lambda k_1 + (1 - \lambda)k_2 = \bar{k} \in K, 0 \leq \lambda \leq 1$ , is also a best simultaneous approximation to  $x_1, x_2$ .*

**Proof.** Since

$$\begin{aligned} & \max(\|x_1 - \bar{k}\|, \|x_2 - \bar{k}\|) \\ &= \max(\|\lambda(x_1 - k_1) + (1 - \lambda)(x_1 - k_2)\|, \|\lambda(x_2 - k_1) + (1 - \lambda)(x_2 - k_2)\|) \\ &\leq \max(\lambda \|x_1 - k_1\| + (1 - \lambda) \|x_1 - k_2\|, \lambda \|x_2 - k_1\| + (1 - \lambda) \|x_2 - k_2\|) \\ &\leq \lambda \max(\|x_1 - k_1\|, \|x_2 - k_1\|) + (1 - \lambda) \max(\|x_1 - k_2\|, \|x_2 - k_2\|) \\ &\leq \lambda d(x_1, x_2; k) + (1 - \lambda) d(x_1, x_2; k) \\ &= d(x_1, x_2; k) \end{aligned}$$

and the reverse inequality always holds, we conclude that:

$$\max(\|x_1 - \bar{k}\|, \|x_2 - \bar{k}\|) = d(x_1, x_2, k).$$

3. If  $K$  is a subspace of a strictly convex normed linear space  $X$ , then it is known that there is at most one best approximation to any element  $x \in X - K$ . In this section we shall prove a similar result for the best simultaneous approximation.

**PROPOSITION 3.1.** *Let  $K$  be a subspace of a strictly convex normed linear space  $X$ . Then there is at most one best simultaneous approximation from the elements of  $K$ , to any two elements  $x_1, x_2 \in X$ .*

**Proof.** Suppose  $k_1$  and  $k_2$  are best simultaneous approximations to  $x_1, x_2$ . Let  $d = \max(\|x_1 - k_i\|, \|x_2 - k_i\|)$ , ( $i = 1, 2$ ). Then there are two cases to consider.

(a) Let  $\|x_1 - k_1\| = d$  and  $\|x_2 - k_1\| = l < d$  (or vice-versa), and write  $d - l = \varepsilon$ . We can find a convex neighbourhood  $U \subset K$  of  $k_1$  such that:

$$d - \varepsilon/4 \leq \|x_1 - k\| \leq d + \varepsilon/4$$

and

$$l - \varepsilon/4 \leq \|x_2 - k\| \leq l + \varepsilon/4, \quad \forall k \in U.$$

Thus  $\max(\|x_1 - k\|, \|x_2 - k\|) = \|x_1 - k\|$  whenever  $k \in U$ . Further,  $\|x_1 - k\| \geq d$ . The element  $\bar{k} = \lambda k_2 + (1 - \lambda)k_1 \in U$  provided  $\lambda$  is sufficiently small and non-zero. Since  $\bar{k}$  is also a best simultaneous approximation by lemma 2.3, we have  $\|x_1 - \bar{k}\| = d$ . However  $\|x_1 - k_1\| = d$  and  $\|x_1 - (k_1 + \bar{k})/2\| = d$ . From these last three relations and the strict convexity of the norm we deduce that  $k_1 = \bar{k}$ , thus  $k_1 = k_2$ .

(b) Assume  $\|x_1 - k_1\| = \|x_2 - k_1\| = d$ , and also  $\|x_1 - k_2\| = \|x_2 - k_2\| = d$  (if not then the previous argument holds). Write:  $\bar{k} = (k_1 + k_2)/2$ , then there are three

possibilities, either

- (i)  $\|x_1 - \bar{k}\| = \|x_2 - \bar{k}\| = d,$
- (ii)  $\|x_1 - \bar{k}\| = d$  and  $\|x_2 - \bar{k}\| < d,$  or
- (iii)  $\|x_1 - \bar{k}\| < d, \|x_2 - \bar{k}\| = d.$

In all the three cases we have either:

$$\begin{aligned} \|x_1 - k_1\| &= \|x_1 - k_2\| = \|x_1 - (k_1 + k_2)/2\|, \text{ or} \\ \|x_2 - k_1\| &= \|x_2 - k_2\| = \|x_2 - (k_1 + k_2)/2\| \end{aligned}$$

or both. Using the strict convexity of the norm we deduce that  $k_1 = k_2.$

4. Let  $K$  be a closed and convex subset of a Banach Space  $X.$  If  $X$  is uniformly convex, then every element in  $X$  has a unique best approximation from the elements of  $K.$  In this section we show that a similar result holds for best simultaneous approximation.

PROPOSITION 4.1. *Let  $K$  and  $X$  be as above, then any two elements  $x_1, x_2 \in X$  have a unique best simultaneous approximation from the elements of  $K.$*

**Proof.** Let

$$d = \inf_{k \in K} \max(\|x_1 - k\|, \|x_2 - k\|)$$

and  $\{k_n\}$  be a sequence of elements in  $K$  such that:

$$\lim_{n \rightarrow \infty} \max(\|x_1 - k_n\|, \|x_2 - k_n\|) \rightarrow d.$$

We can assume without loss of generality that  $d=1.$

Let  $d_m = \max(\|x_1 - k_m\|, \|x_2 - k_m\|),$  then  $d_m \geq 1$  and

$$(4.1) \quad \frac{\|x_i - k_m\|}{d_m} \leq 1.$$

Consider

$$\frac{1}{2} \left( \frac{k_m}{d_m} + \frac{k_n}{d_n} \right) = \frac{d_n k_m + d_m k_n}{d_m + d_n} \cdot \frac{d_m + d_n}{2 d_m d_n}$$

and write

$$y_{mn} = \frac{d_n k_m + d_m k_n}{d_m + d_n}.$$

Since  $K$  is convex  $y_{mn} \in K.$  Hence  $\max(\|x_1 - y_{mn}\|, \|x_2 - y_{mn}\|) \geq 1$  and consequently

$$\begin{aligned} \max \left( \left\| \frac{d_m + d_n}{2 d_m d_n} x_1 - \frac{1}{2} \left( \frac{k_m}{d_m} + \frac{k_n}{d_n} \right) \right\|, \left\| \frac{d_m + d_n}{2 d_m d_n} x_2 - \frac{1}{2} \left( \frac{k_m}{d_m} + \frac{k_n}{d_n} \right) \right\| \right) \\ = \max(\|x_1 - y_{mn}\|, \|x_2 - y_{mn}\|) \frac{d_m + d_n}{2 d_m d_n} \geq \frac{d_m + d_n}{2 d_m d_n}. \end{aligned}$$

Therefore at least one of the following is true:

$$(4.2) \quad \left\| \frac{x_1 - k_m}{d_m} + \frac{x_1 - k_n}{d_n} \right\| \geq \frac{d_m + d_n}{d_m d_n}$$

$$(4.3) \quad \left\| \frac{x_2 - k_m}{d_m} + \frac{x_2 - k_n}{d_n} \right\| \geq \frac{d_m + d_n}{d_m d_n}.$$

Suppose (4.2) is true, then, from (4.1) and the uniform convexity of the norm it follows that for any given  $\varepsilon > 0$ , there exists a  $N$  such that

$$(4.4) \quad \left\| \frac{x_1 - k_m}{d_m} - \frac{x_1 - k_n}{d_n} \right\| < \varepsilon \quad \text{for } m, n > N.$$

Using (4.4) and the fact that  $d_m \rightarrow 1$  it can be shown that the sequence  $\{k_n\}$  is a Cauchy sequence, hence it converges to some  $k$  in  $X$ . Since  $K$  is closed,  $k \in K$ . The element  $k$  is the unique best simultaneous approximation.

5. In an inner produce space the problem of best simultaneous approximation is relatively much easier. Let  $H$  be a real inner produce space and  $G$  a subspace of  $H$ . Consider two elements  $x_1, x_2 \in H$ , which have best approximations, say  $g_1, g_2$  from the elements of  $G$ . If  $\|x_1 - g_2\| \leq \|x_2 - g_2\|$ , then  $g_2$  is also a best simultaneous approximation. Similarly if  $\|x_2 - g_1\| \leq \|x_1 - g_1\|$ , then  $g_1$  is also a best simultaneous approximation. If the above two conditions are not satisfied then

$$\bar{g} = \lambda g_1 + (1 - \lambda)g_2, \quad (0 < \lambda < 1),$$

is the best simultaneous approximation, where  $\lambda$  is given by

$$(5.1) \quad \|x_1 - \bar{g}\| = \|x_2 - \bar{g}\|.$$

For this we need to show that

$$\max(\|x_1 - \bar{g} + g\|, \|x_2 - \bar{g} + g\|) \geq \|x_1 - \bar{g}\| = (\|x_2 - \bar{g}\|) \quad \forall g \in G.$$

On the contrary suppose that there exists a  $g \in G$  such that

$$(5.2) \quad \|x_1 - \bar{g} + g\| < \|x_1 - \bar{g}\|, \quad \text{and}$$

$$(5.3) \quad \|x_2 - \bar{g} + g\| < \|x_1 - \bar{g}\|.$$

From (5.2) and (5.3) we obtain

$$(x_1 - g_2, g) < -\frac{(g, g)}{2(1 - \lambda)} \quad \text{and}$$

$$(x_2 - g_1, g) < -\frac{(g, g)}{2\lambda}.$$

Adding these two we get

$$(x_1 - g_2 + x_2 - g_1, g) < -\frac{(g, g)}{2} \left[ \frac{1}{\lambda} + \frac{1}{1-\lambda} \right]$$

or

$$(x_1 - g_1, g) + (x_2 - g_2, g) < -\frac{(g, g)}{2} \left[ \frac{1}{\lambda} + \frac{1}{1-\lambda} \right]$$

i.e.  $0 < -(g, g)/2[(1/(1-\lambda)) + 1/\lambda]$  since by hypothesis  $x_1 - g_1, x_2 - g_2 \perp G$  which is a contradiction.

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