

## Duality in Nonlinear Approximation\*

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### 1. INTRODUCTION

We are mainly concerned with the problem of characterizing the distance  $d_x(W)$  between a fixed point  $x$  and a nonempty subset  $W$  of a real normed linear space  $X$ . We intend to develop a purely geometric concept by which  $d_x(W)$  can be estimated from below and even obtained as the maximum of certain lower bounds.

In Section 2 we start with the consideration of families  $\mathcal{R}$  of half spaces  $R$  in  $X$  such that

$$W \subseteq \bigcup_{R \in \mathcal{R}} R \quad \text{and} \quad x \notin \bigcup_{R \in \mathcal{R}} R.$$

By  $\mathcal{H}$  we denote the family of corresponding hyperplanes  $H$ . It is geometrically evident that the infimum of all the distances  $d_x(H)$  from  $x$  to  $H \in \mathcal{H}$  is a lower bound for  $d_x(W)$ . This is proved as Lemma 2.2.

The main result of Section 2 is a duality theorem which states that  $d_x(W)$  is the maximum of all such infima. This generalizes the well-known fact that, if  $W$  is convex,  $d_x(W)$  is the maximum of all the distances  $d_x(H)$  where  $H$  is a hyperplane separating  $x$  and  $W$  (Theorem 2.5).

In Section 3 we introduce supporting systems and strong supporting systems for  $W$ . The latter play the major role since they serve as an important tool in the characterization of projection points  $\hat{w} \in W$ , i.e., points  $\hat{w}$  such that  $\|\hat{w} - x\| = d_x(W)$ .

A strong supporting system for  $W$  is a family  $\mathcal{R}$  of half spaces  $R$  such that

$$W \subseteq \bigcup_{R \in \mathcal{R}} R \quad \text{and} \quad S = \bigcap_{H \in \mathcal{H}} \{H \cap W\}$$

is nonempty, where  $\mathcal{H}$  is the family of corresponding hyperplanes. The elements of  $S$  are called supporting points. For instance, if  $W$  is convex, each supporting hyperplane defines a strong supporting system which consists of only one half space.

In Theorem 3.3 we obtain a well-known sufficient condition for a point  $\hat{w} \in W$  to be a projection point. If  $W$  is convex then a restricted form of this

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condition is also necessary. The formulation we give makes use of supporting systems and thus yields a purely geometric viewpoint.

We conclude Section 3 with the consideration of the following situation which generalizes several special cases: Let  $Y$  be another real normed linear space and  $A$  a nonempty open subset of  $Y$ . We consider a Fréchet-differentiable mapping  $F: A \rightarrow X$  and put  $W = F(A)$ . We have investigated this case in [11] and we give a short review of the results at the end of Section 3.

At the beginning of Section 4 we present an algebraic version of Lemma 2.2. Then we consider the special case of the Chebychev approximation problem where  $X$  is the vector space  $C(M)$  of all continuous, real valued functions on a compact Hausdorff space  $M$  with the maximum norm. The concept of  $H$ -sets in  $M$ , due to Collatz [2], can be formulated in terms of strong supporting systems. Furthermore, a result of Collatz concerning lower bounds for  $d_x(W)$  and a similar one of Meinardus and Schwedt [12] turn out to be special cases of Lemma 2.2. The case  $W = F(A)$  where  $A$  is a nonempty open subset of the real euclidean  $n$ -space and  $F: A \rightarrow C(M)$  is a Fréchet-differentiable mapping has been investigated in [10], so that we content ourselves with a short review of the results. Finally, we treat the case of discrete  $L_p$ -approximation and give a simple method to verify the assumptions of Lemma 2.2 for the generalized rational approximation problem.

## 2. A DUALITY THEOREM

We consider a normed linear space  $X$  over the reals and denote the norm by  $\|\cdot\|$ . Let  $X^*$  be the dual space of  $X$ , that is, the set of all continuous linear functionals  $L$  mapping  $X$  into the reals.  $X^*$  becomes a real Banach space if we define the norm by

$$\|L\| = \sup_{\|x\|=1} |L(x)|, \quad L \in X^*.$$

By  $S^*$  we denote the unit sphere of  $X^*$ , that is, the set of all  $L \in X^*$  such that  $\|L\| = 1$ . In the following, a half space  $R$  of  $X$  is always defined by a pair  $(L, \alpha)$  with  $L \in S^*$  and  $\alpha$  real, so that

$$R = \{y \in X : L(y) \geq \alpha\}, \tag{2.1}$$

We call

$$H = \{h \in X : L(h) = \alpha\}$$

the corresponding hyperplane.

LEMMA 2.1 [3]. *For a given  $L \in S^*$ , let the half space  $R$  be defined by (2.1) and let  $H$  be the corresponding hyperplane. Then for each  $x \notin R$ , the distance*

$$d_x(H) = \inf_{h \in H} \|h - x\|$$

from  $x$  to  $H$  is given by

$$d_x(H) = L(h - x) = \alpha - L(x)$$

for all  $h \in H$ .

*Proof.*  $x \notin R$  implies  $x \notin H$ . Since  $H$  is closed we have  $d_x(H) > 0$ . By assumption the closed ball

$$K = \{y \in X : \|y - x\| \leq d_x(H)\}$$

is contained in the half space

$$\{y \in X : L(y) \leq \alpha\}.$$

This implies

$$L(h) = \alpha \geq \sup_{y \in K} L(y) = d_x(H) + L(x)$$

or

$$\alpha - L(x) = L(h - x) \geq d_x(H) \quad \text{for all } h \in H.$$

On the other hand we have for each  $h \in H$

$$\alpha - L(x) = L(h - x) \leq \|h - x\|,$$

and therefore

$$L(h - x) = \alpha - L(x) \leq d_x(H).$$

This completes the proof.

**LEMMA 2.2.** *Let  $\mathcal{R}$  be a family of half spaces and  $\mathcal{H}$  the family of corresponding hyperplanes. For a nonempty subset  $W$  of  $X$  and for an  $x \in X$  we assume*

$$W \subseteq \bigcup_{R \in \mathcal{R}} R \tag{2.2}$$

and

$$x \notin \bigcup_{R \in \mathcal{R}} R. \tag{2.3}$$

Then we have

$$\inf_{H \in \mathcal{H}} d_x(H) \leq d_x(W) = \inf_{w \in W} \|w - x\|. \tag{2.4}$$

*Proof.* Every  $R \in \mathcal{R}$  is given by (2.1) for some  $L \in S^*$ . Let  $\mathcal{L}$  be the collection of all these  $L$ . (2.2) then implies that for each  $w \in W$  there exists  $H \in \mathcal{H}$  and  $L \in \mathcal{L}$  such that

$$L(w) \geq L(h) \quad \text{for all } h \in H. \tag{2.5}$$

(2.3) implies, by Lemma 2.1, that for each  $R \in \mathcal{R}$

$$d_x(H) = L(h - x) \quad \text{for all } h \in H,$$

where  $H$  is the corresponding hyperplane and  $L$  the corresponding element of  $\mathcal{L}$ .

Suppose (2.4) is false. Then by the definition of  $d_x(W)$  there is a  $\hat{w} \in W$  such that

$$d_x(H) > \|\hat{w} - x\| \quad \text{for all } H \in \mathcal{H}.$$

Hence for each  $H \in \mathcal{H}$  and the corresponding  $L \in \mathcal{L}$  we have

$$\begin{aligned} L(h - \hat{w}) &= L(h - x) - L(\hat{w} - x) \geq L(h - x) - \|\hat{w} - x\| \\ &= d_x(H) - \|\hat{w} - x\| > 0 \quad \text{for all } h \in H. \end{aligned}$$

This contradicts (2.5). Therefore (2.4) must be true.

Now we assume  $W$  to be a nonempty convex subset of  $X$  and  $x \in X$  to be a point not belonging to the closure of  $W$ . According to a well-known separation theorem [9] there is a half space  $R$  such that  $W \subseteq R$  and  $x \notin R$ . By Lemma 2.2 we therefore have  $d_x(H) \leq d_x(W)$  where  $H$  is the corresponding hyperplane.

LEMMA 2.3. *In addition to the assumptions of Lemma 2.2 we require  $W \subseteq X$  to be convex and*

$$d = \inf_{H \in \mathcal{H}} d_x(H) > 0.$$

*Then there is a half space  $\hat{R}$  such that  $W \subseteq \hat{R}$ ,  $x \notin \hat{R}$  and*

$$d \leq d_x(\hat{H}) \leq d_x(W), \tag{2.6}$$

*where  $\hat{H}$  is the corresponding hyperplane.*

*Proof.* By the above separation theorem [9] the closed ball

$$K_d = \{y \in X : \|y - x\| \leq d\}$$

and  $W$  can be separated by a hyperplane

$$\hat{H} = \{h \in X : \hat{L}(h) = \hat{\alpha}\},$$

where  $\hat{L} \in S^*$  and  $\hat{\alpha}$  is a real scalar; i.e.,

$$W \subseteq \hat{R} = \{y \in X : \hat{L}(y) \geq \hat{\alpha}\} \quad \text{and} \quad K_d \subseteq \{y \in X : \hat{L}(y) \leq \hat{\alpha}\};$$

in particular,  $x \notin \hat{R}$ . This implies

$$\inf_{w \in W} \hat{L}(w) \geq \hat{\alpha} \geq \sup_{y \in K_d} \hat{L}(y) = d + \hat{L}(x),$$

and applying Lemma 2.1 we conclude

$$d_x(W) \geq \inf_{w \in W} \hat{L}(w - x) \geq \hat{\alpha} - L(x) = d_x(\hat{H}) \geq d$$

which completes the proof.

LEMMA 2.4. *Let  $x \in X$  and a nonempty subset  $W$  of  $X$  be given such that*

$$d_x(W) = \inf_{w \in W} \|w - x\| > 0.$$

Then there exists a family  $\mathcal{R}$  of half spaces such that (2.2) and (2.3) hold and

$$d_x(H) = d_x(W) \quad (2.7)$$

for all  $H$  of the family  $\mathcal{H}$  of corresponding hyperplanes.

*Proof.* Put  $K = \{y \in X: \|y - x\| \leq d_x(W)\}$ . If we choose an arbitrary  $w \in W$  and define

$$z = (1 - \lambda)x + \lambda w \quad (2.8)$$

where

$$\lambda = \frac{d_x(W)}{\|w - x\|},$$

then

$$\|z - x\| = \lambda\|w - x\| = d_x(W).$$

Furthermore, there exists a hyperplane supporting  $K$  at  $z$  [9], which is given by

$$H_z = \{h \in X: L_z(h) = L_z(z)\}$$

where  $L_z \in S^*$ . With no loss of generality we may assume

$$L_z(x) < L_z(z) = L_z(h) \quad \text{for all } h \in H_z. \quad (2.9)$$

On the other hand, we have

$$L_z(w - z) = \frac{1 - \lambda}{\lambda} L_z(z - x) \geq 0. \quad (2.10)$$

If we define

$$R_z = \{y \in X: L_z(y) \geq L_z(z)\}$$

and denote by  $\mathcal{R}$  the family of all such half spaces  $R_z$  where  $z$  is defined by (2.8) and  $w$  varies over  $W$ , then (2.2) and (2.3) are an immediate consequence of (2.9) and (2.10). Furthermore, we have

$$L_z(z) \geq \sup_{y \in K} L_z(y) = d_x(W) + L_z(x),$$

whence by Lemma 2.1

$$d_x(H_z) = L_z(z - x) \geq d_x(W).$$

On the other hand

$$d_x(H_z) = L_z(z - x) \leq \|z - x\| = d_x(W).$$

This completes the proof.

Lemma 2.2 and 2.3 yield the following

**DUALITY THEOREM.** *If for  $x \in X$  and a nonempty subset  $W$  of  $X$  we have  $d_x(W) > 0$ , then*

$$d_x(W) = \max_{\mathcal{R}} \inf_{H \in \mathcal{H}} d_x(H),$$

where the maximum is taken over all families  $\mathcal{R}$  of half spaces satisfying (2.2) and (2.3), and  $\mathcal{H}$  is the family of corresponding hyperplanes.

From now on we assume  $W$  to be convex. If for some  $x \in X$  we have  $d_x(W) > 0$ , then by the Duality Theorem there is a family  $\mathcal{R}$  of half spaces satisfying (2.2) and (2.3) such that

$$d_x(W) = \inf_{H \in \mathcal{H}} d_x(H) > 0,$$

where  $\mathcal{H}$  is the corresponding family of hyperplanes. However, by Lemma 2.3 there exists a half space  $\hat{R}$  such that  $W \subseteq \hat{R}$ ,  $x \notin \hat{R}$ , and

$$d_x(\hat{H}) = \inf_{H \in \mathcal{H}} d_x(H)$$

where  $\hat{H}$  is the corresponding hyperplane.

Using this result and Lemma 2.2 we get

**THEOREM 2.5.** *Let  $W$  be a nonempty convex subset of  $X$  and assume  $d_x(W) > 0$  for some  $x \in X$ . Then*

$$d_x(W) = \max_R d_x(H)$$

where the maximum is taken over all the half spaces  $R$  such that  $W \subseteq R$ ,  $x \notin R$ , and  $H$  is the corresponding hyperplane.

This result is well known (compare, for instance, [3], [5], [7], where equivalent results are obtained) and can now be considered as a special case of the above Duality Theorem.

### 3. SUPPORTING SYSTEMS AND A SUFFICIENT CONDITION FOR PROJECTION POINTS

As in Section 2, we start with a real normed linear space  $X$  and consider a family  $\mathcal{R}$  of half spaces  $R$  defined by (2.1), where  $L \in \mathcal{L}$ , and  $\mathcal{L}$  is the corresponding set of linear forms in  $S^*$ .

**DEFINITION.**  $\mathcal{R}$  is called a supporting system for a nonempty subset  $W$  of  $X$  if (2.2) holds and if for all  $H \in \mathcal{H}$  we have

$$H \cap W \neq \emptyset, \tag{3.1}$$

where  $\mathcal{H}$  is the corresponding family of hyperplanes and  $\emptyset$  denotes the empty set.

If  $\mathcal{R}$  is a supporting system for the nonempty subset  $W$  of  $X$  and  $x \in X$  is such that (2.3) holds then Lemma 2.2 yields a lower bound for the distance

$$d_x(W) = \inf_{w \in W} \|w - x\|$$

between  $x$  and  $W$ .

**LEMMA 3.1.** *If for  $x \in X$  and for a nonempty subset  $W$  of  $X$  we have  $d_x(W) > 0$ , then there is a supporting system  $\mathcal{R}$  for  $W$  such that (2.3) is satisfied.*

*Proof.* If in the proof of Lemma 2.4 we substitute the family  $\mathcal{R} = \{R_z\}$  by  $\hat{\mathcal{R}} = \{\hat{R}_z\}$ , where

$$\hat{R}_z = \{y \in X : L_z(y) \geq L_z(w)\}$$

and  $z$  is given by (2.8), then it is easy to verify that  $\hat{\mathcal{R}}$  satisfies (2.2), (2.3), and (3.1).

**DEFINITION.** A supporting system  $\mathcal{R}$  for the nonempty subset  $W$  of  $X$  is called a strong supporting system if

$$S = \bigcap_{H \in \mathcal{R}} \{H \cap W\} \neq \emptyset, \quad (3.2)$$

where  $\mathcal{R}$  is the corresponding family of hyperplanes.

The elements of  $S$  are called supporting points.

Now let  $\mathcal{R}$  be a strong supporting system for the nonempty subset  $W$  of  $X$  and let  $\hat{w} \in W$  be an arbitrary, but fixed, supporting point. Then each  $R \in \mathcal{R}$  is of the form

$$R = R_L = \{y \in X : L(y) \geq L(\hat{w})\} \quad (3.3)$$

where  $L \in \mathcal{L}$ .

The condition (2.2) is therefore equivalent to the following: For each  $w \in W$  there is an  $L \in \mathcal{L}$  such that

$$L(w) \geq L(\hat{w}). \quad (3.4)$$

The condition (2.3) is equivalent to

$$L(x) < L(\hat{w}) \quad (3.5)$$

for all  $L \in \mathcal{L}$ . By Lemmas 2.1 and 2.2 the inequalities (3.4) and (3.5) imply

$$\inf_{L \in \mathcal{L}} L(\hat{w} - x) \leq d_x(W). \quad (3.6)$$

If we, furthermore, assume that  $\mathcal{L}$  is a nonempty weakly\* closed subset of  $S^*$ , hence weakly\* compact, then (3.4) is equivalent to

$$\min_{L \in \mathcal{L}} L(\hat{w} - w) \leq 0 \quad (3.7)$$

for all  $w \in W$ , and in (3.6) "inf" can be replaced by "min."

LEMMA 3.2. *Let  $\mathcal{R}$  be a family of half spaces  $R_L$  defined by (3.3), where  $L \in \mathcal{L}$ , and  $\mathcal{L}$  is a nonempty weakly\* closed subset of  $S^*$ . Let  $E(\mathcal{L})$  be the (nonempty [9]) set of extreme points of  $\mathcal{L}$ , and  $\tilde{\mathcal{R}}$  the family of all  $R_L \in \mathcal{R}$  such that the corresponding  $L$  is an element of  $E(\mathcal{L})$ . If  $\mathcal{R}$  is a strong supporting system for a nonempty subset  $W$  of  $X$  then the same is true for  $\tilde{\mathcal{R}}$ .*

*Proof* (as in [8]). Let  $\hat{w} \in W$  be an arbitrary supporting point. For each  $w \in W$  we define a linear functional  $g_w$  mapping  $\mathcal{L}$  into the reals by  $g_w(L) = L(w - \hat{w})$  where  $L \in \mathcal{L}$ . As  $g_w$  is weakly\* continuous,  $g_w(\mathcal{L})$  is a compact subset of the reals and, therefore, has an extreme point  $r \geq 0$ , since  $\mathcal{R}$  is a strong supporting system. It is well known [9] that  $r$  is the image of an extreme point  $L_e \in E(\mathcal{L})$ . Hence, for each  $w \in W$ , there exists an  $L_e \in E(\mathcal{L})$  such that  $L_e(w - \hat{w}) \geq 0$ , which completes the proof.

Let  $x \in X$  be a fixed point, and  $W$  a nonempty subset of  $X$ .  $\hat{w} \in W$  is called a projection point of  $x$  in  $W$  if

$$\|\hat{w} - x\| = d_x(W).$$

In the following, we assume  $d_x(W) > 0$ . For each  $w \in W$  we define a set:

$$E_w = \{L \in S^* : L(w - x) = \|w - x\|\}.$$

By the Hahn–Banach Theorem,  $E_w$  is nonempty, and, furthermore,  $E_w$  is obviously a weakly\* closed (hence weakly\* compact) convex subset of  $S^*$ .

THEOREM 3.3. *For some  $\hat{w} \in W$ , let  $\tilde{\mathcal{R}}$  be the family of half spaces  $R_L$  defined by (3.3), where  $L \in \mathcal{L}$ , and  $\mathcal{L}$  is a nonempty subset of  $E_{\hat{w}}$ . If  $\tilde{\mathcal{R}}$  is a strong supporting system for  $W$ , then  $\hat{w}$  is a projection point of  $x$  in  $W$ .*

*Proof.* By assumption, (3.4) and (3.5) are satisfied, the latter because of

$$0 < d_x(W) \leq \|\hat{w} - x\| = L(\hat{w} - x) \quad \text{for all } L \in \mathcal{L}.$$

Hence (3.6) holds, implying

$$\|\hat{w} - x\| = \inf_{L \in \mathcal{L}} L(\hat{w} - x) \leq d_x(W).$$

If  $\mathcal{L}$  is a nonempty weakly\* closed subset of  $E_w$ , then  $\mathcal{L}$  is also weakly\* closed in  $S^*$ , and therefore the assumption of Theorem 3.3 is equivalent to (3.7). By Lemma 3.2, the assumption of Theorem 3.3 remains true if we replace  $\mathcal{L}$  by the set  $E(\mathcal{L})$  of its extreme points.

For applications it is important to know whether, in this case,  $E(\mathcal{L})$  is contained in the set  $E(K^*)$  of extreme points of the unit ball  $K^*$  of  $X^*$ , because in various special cases,  $E(K^*)$  has a rather simple structure.



The statement of Theorem 3.3 is not new. In [13] Nikolskii considers the case where  $\mathcal{L}$  is the intersection of  $E_{\hat{w}}$  and a so-called fundamental system  $\Gamma$ , which is a weakly\* closed subset of  $K^*$  such that for each  $y \neq 0$  in  $X$  there exists an  $L \in \Gamma$  with  $L(y) = \|y\|$ . Examples of fundamental systems are the unit sphere  $S^*$  and the closure of  $E(K^*)$ .

Nikolskii proves that  $W \subseteq \bigcup_{L \in E(\mathcal{L})} R_L$  is a sufficient condition for  $\hat{w} \in W$  to be a projection point of  $x$  in  $W$ , and a necessary condition in the case where  $W$  is convex.

In [6] Garkavi obtains the same result as Nikolskii, with the only differences that instead of  $E(\mathcal{L})$ , the set  $E(E_{\hat{w}}) = E_{\hat{w}} \cap E(K^*)$  of extreme points of  $E_{\hat{w}}$  is considered and that  $X$  is a Banach space. Recently, Deutsch and Maserick [3] reproved this result for a normed linear space  $X$ .

In [7], Havinson gives the same characterization for projection points in convex sets as Garkavi. Furthermore, he obtains the following criterion which is a simple consequence of Theorem 2.5: If  $W$  is a nonempty convex subset of  $X$ , and  $x \in X$  is such that  $d_x(W) > 0$ , then  $\hat{w} \in W$  is a projection point of  $x$  in  $W$  if and only if there exists an element  $L \in S^*$  such that  $L \in E_{\hat{w}}$  and  $L(\hat{w}) \leq L(w)$  for all  $w \in W$ .

However, as Deutsch and Maserick point out in [3], this  $L$  cannot, in general, be chosen to be an element of  $E(K^*)$ .

Brosowski considers the case  $\mathcal{L} = E_{\hat{w}} \cap \Gamma$ , where  $\Gamma$  is a fundamental system, and studies the question: For what nonempty subsets  $W$  of  $X$  other than convex subsets is the condition  $W \subseteq \bigcup_{L \in E(\mathcal{L})} R_L$  necessary for  $\hat{w}$  to be a projection point of  $x$  in  $W$ ? He states that the condition is necessary for so-called  $\Gamma$ -regular subsets of  $X$ . For details we refer to [1], where the results are given without proofs. These are to appear in a forthcoming paper.

In [11] we have investigated the following situation which occurs in various special cases: Let  $Y$  be a real normed linear space,  $A$  a nonempty open subset of  $Y$ , and  $F: A \rightarrow X$  a mapping such that for each  $a \in A$  the Fréchet derivative  $F'_a$  exists. For  $W$  we take the image  $F(A)$ , and we consider an element  $x \in X$  such that

$$d_x(W) = \inf_{a \in A} \|F(a) - x\| > 0.$$

We then obtain the following necessary condition for a projection point; we assume that for every fixed  $h \in Y$  the mapping  $a \rightarrow F'_a(h)$ ,  $a \in A$ , is continuous. If  $F(\hat{a})$ ,  $\hat{a} \in A$ , is a projection point of  $x$  in  $W$ , then for each  $h \in Y$  there exists an  $L \in E_{F(\hat{a})}$  such that

$$L(F'_a(h)) \leq 0. \tag{3.8}$$

This result has also been given by Henze in [8], however, without the above continuity assumption on the mappings  $a \rightarrow F'_a(h)$ . But this is indispensable.

If we define for each  $a \in A$  the linear manifold

$$T_a = \{F(a) - F'_a(h) : h \in Y\}$$

and consider the family  $\mathcal{R}$  of all half spaces

$$R_L = \{y \in X : L(y) \geq L(F(\hat{a}))\}, \quad L \in E_{F(\hat{a})},$$

then condition (3.8) is equivalent to  $\mathcal{R}$  being a strong supporting system for  $T_{\hat{a}}$ , with  $F(\hat{a})$  as supporting point.

Furthermore, Henze shows in [8] that condition (3.8) remains true if we take the set  $E(E_{F(\hat{a})})$  of extreme points of  $E_{F(\hat{a})}$  instead of  $E_{F(\hat{a})}$ . This is also an immediate consequence of Lemma 3.2.

In order to prove that condition (3.8) with  $E(E_{F(\hat{a})})$  instead of  $E_{F(\hat{a})}$  is sufficient for  $F(\hat{a})$  to be a projection point of  $x$  in  $F(A)$ , we assumed in [11] that  $F$  has the following property: For each pair  $(a, b) \in A \times A$ , there is a positive continuous functional  $\phi_{a,b} : E(K^*) \rightarrow \mathfrak{R}$  and an element  $h = h(a, b) \in Y$  such that

$$L(F(a) - F(b)) = \phi_{a,b}(L) \cdot L(F'_a(h)). \tag{3.9}$$

In the case of the Chebychev approximation problem (compare Section 4) this property is essentially equivalent to the asymptotic convexity of  $F$  introduced by Meinardus and Schwedt in [12]. This was shown in [10]. We therefore call  $F$  asymptotically convex if it has the property (3.9).

Under the assumptions that  $F$  is asymptotically convex and that for each fixed  $h \in Y$  the mapping  $a \rightarrow F'_a(h)$ ,  $a \in A$ , is continuous, we have shown in [11] that for  $F(\hat{a})$ ,  $\hat{a} \in A$ , to be a projection point of  $x$  in  $W = F(A)$ , the following condition is necessary: For each  $a \in A$  there is an  $L \in E(E_{F(\hat{a})})$  such that

$$L(F(a)) \geq L(F(\hat{a})). \tag{3.10}$$

Finally, we give a somewhat negative result which is also contained in [11]. We assume  $X$  to be flat convex [9], that is, at each point of the unit sphere of  $X$  there is exactly one supporting hyperplane of the unit ball. Examples of flat convex normed linear spaces are Hilbert spaces and  $L_p$ -spaces with  $1 < p < \infty$ .

If, furthermore,  $F : A \rightarrow X$  is asymptotically convex, if for each fixed  $h \in Y$  the mapping  $a \rightarrow F'_a(h)$ ,  $a \in A$ , is continuous, and if for each  $x \in X$  there exists a projection point in  $W = F(A)$ , then  $W$  is a linear manifold.

#### 4. SPECIAL CASES AND EXAMPLES

We start with an algebraic version of Lemma 2.2. Let  $X$  be a real normed linear space,  $X^*$  its dual, and  $\mathcal{L}$  a nonempty subset of the unit sphere  $S^*$  of  $X^*$ . To each  $L \in \mathcal{L}$  we assign a real number  $\alpha_L$ . Then we have the following:

Let  $W$  be a nonempty subset of  $X$ , and let  $x$  be an arbitrary point of  $X$ . If for each  $w \in W$  there is an  $L \in \mathcal{L}$  such that

$$L(w) \geq \alpha_L, \quad (4.1)$$

then

$$\inf_{L \in \mathcal{L}} \{\alpha_L - L(x)\} \leq d_x(W). \quad (4.2)$$

If  $L(x) \geq \alpha_L$  for some  $L \in \mathcal{L}$ , the assertion (4.2) is trivial. We therefore assume

$$L(x) < \alpha_L \quad \text{for all } L \in \mathcal{L}. \quad (4.3)$$

We put for each  $L \in \mathcal{L}$

$$R_L = \{y \in X : L(y) \geq \alpha_L\} \quad (4.4)$$

and define

$$\mathcal{R} = \{R_L : L \in \mathcal{L}\}.$$

Then (4.1) and (4.3) are equivalent to (2.2) and (2.3), thus implying (4.2) by Lemmas 2.1 and 2.2.

Now we consider special cases.

(a) Uniform approximation: Let  $X$  be the vector space  $C(M)$  of real valued continuous functions defined on a compact Hausdorff space  $M$ . The norm in  $X = C(M)$  will be the maximum norm

$$\|g\| = \max_{P \in M} |g(P)|, \quad g \in C(M).$$

Let  $W$  be a nonempty subset of  $C(M)$  and let  $f$  be a point of  $C(M)$  not belonging to the closure of  $W$ .

The problem of finding projection points  $\hat{w} \in W$  of  $f$  in  $W$  is the well-known nonlinear Chebychev approximation problem.

For  $X = C(M)$ , the set of extreme points  $E(K^*)$  of the unit ball  $K^*$  of  $X^*$ , is given by

$$E(K^*) = \{\epsilon_P \delta_P : P \in M, \epsilon_P = +1 \text{ or } -1\},$$

where  $\delta_P$  is the point measure in  $P$ , i.e.,

$$\delta_P(g) = g(P) \quad \text{for all } g \in C(M).$$

Let  $D$  be a nonempty subset of  $M$ . To each  $P \in D$  we assign a number  $\epsilon_P \in \{-1, +1\}$  and define  $\mathcal{L}$  by

$$\mathcal{L} = \{\epsilon_P \delta_P : P \in D\}. \quad (4.5)$$

A family  $\mathcal{R}$  of half spaces  $R_L$  of the form (4.4) with  $L \in \mathcal{L}$  is then given by

$$\mathcal{R} = \{R_P : P \in D\} \quad (4.6)$$

where

$$R_P = \{g \in C(M) : \epsilon_P g(P) \geq \alpha_P\} \quad (4.7)$$

and  $\alpha_P$  is a real scalar assigned to  $P$ .

Collatz [2] calls  $D$  an  $H$ -set if  $D$  is the disjoint union of two nonempty sets  $D_1$  and  $D_2$  such that for no pair  $w, \hat{w} \in W$  it is true that

$$w(P) - \hat{w}(P) \begin{cases} < 0 & \text{for all } P \in D_1 \\ > 0 & \text{for all } P \in D_2. \end{cases} \tag{4.8}$$

Assume  $D$  to be an  $H$ -set and consider an arbitrary but fixed  $\hat{w} \in W$ . Then (4.8) implies that for each  $w \in W$  there is a  $P \in D$  with

$$\epsilon_P(w(P) - \hat{w}(P)) \geq 0$$

where

$$\epsilon_P = \begin{cases} +1 & \text{for } P \in D_1 \\ -1 & \text{for } P \in D_2. \end{cases}$$

Defining  $\alpha_P = \epsilon_P \hat{w}(P)$ ,  $P \in D$ , and  $\mathcal{R}$  by (4.6), (4.7), we get the result that  $\mathcal{R}$  is a strong supporting system for  $w$  with  $\hat{w}$  as supporting point.

Besides (4.8), we assume that for some fixed  $\hat{w} \in W$ , we have

$$\epsilon_P(\hat{w}(P) - f(P)) > 0 \quad \text{for all } P \in D.$$

Then it follows that

$$\inf_{P \in D} |\hat{w}(P) - f(P)| \leq d_r(W).$$

This is the contents of Theorem 2 of [2] and a special case of the fact that (3.4) and (3.5) imply (3.6). In [2], Collatz gives various examples of  $H$ -sets and develops a method by which  $H$ -sets can be systematically constructed for certain subsets  $W$  of  $C(M)$ .

Let  $M$  be a finite closed interval  $[a, b]$ . Then, for example, in the case of rational or exponential approximation,  $W$  has the following property: There is a number  $r$  such that no difference  $w - \hat{w}$  of functions  $w, \hat{w} \in W$  has more than  $r$  zeros in  $[a, b]$ . In this case, obviously, each set of  $r + 2$  points  $P_i \in [a, b]$ , with  $a \leq P_1 < \dots < P_{r+2} \leq b$ , is an  $H$ -set.

Now we consider the following situation: Let  $Y$  be a normed linear space and  $A$  a nonempty subset of  $Y$ . Let  $F: A \rightarrow C(M)$  be a given map, and put  $W = F(A)$ . We require  $D$  to be a nonempty closed subset of  $M$  and assume that for some  $\hat{a} \in A$  the following two conditions are satisfied:

$$\min_{P \in D} (F(\hat{a}, P) - f(P))(F(\hat{a}, P) - F(a, P)) \leq 0$$

for all  $a \in A$ , and

$$|F(\hat{a}, P) - f(P)| > 0 \quad \text{for all } P \in D.$$

We define  $\epsilon_P = \text{sgn}(F(\hat{a}, P) - f(P))$ , where  $P \in D$ , and  $\mathcal{L}$  by (4.5). Then (3.4) and (3.5) are satisfied, implying (3.6) with “min” instead of “inf”, since

$\mathcal{L}$  is a weakly\* closed subset of  $K^*$  and hence weakly\* compact. (3.6), in turn, is equivalent to

$$\min_{P \in D} |F(\hat{a}, P) - f(P)| \leq d_f(W).$$

This is exactly Theorem 1 of [12], for the case of real valued functions.

In [10] we considered the case where  $Y$  is the real euclidean space  $\mathfrak{R}^n$  and  $A$  is a nonempty open subset of  $Y$ . If  $F$  is Fréchet-differentiable, then for each  $P \in M$  and  $a = (a_1, \dots, a_n) \in A$ , there exist the partial derivatives

$$\frac{\partial F}{\partial a_j}(a, P), \quad j = 1, \dots, n,$$

and we have

$$F_a'(h) = \sum_{j=1}^n h_j \frac{\partial F}{\partial a_j}(a),$$

where  $h = (h_1, \dots, h_n) \in Y = \mathfrak{R}^n$ .

We have shown in [10] that for each fixed  $h \in Y$ , the mapping  $a \rightarrow F_a'(h)$ ,  $a \in A$ , is continuous if and only if the partial derivatives depend continuously on  $(a, P) \in A \times M$ .

Condition (3.9) immediately leads to the following property of  $F$  (compare condition (5) in [10]): For each pair  $(a, b) \in A \times A$ , there is a positive function  $\phi(a, b) \in C(M)$  and an element  $h = h(a, b) \in \mathfrak{R}^n$  such that

$$F(a) - F(b) = \phi(a, b) \sum_{j=1}^n h_j \frac{\partial F}{\partial a_j}(a).$$

As to the relationship with the asymptotic convexity of  $F$ , introduced by Meinardus and Schwedt in [12], and the discussion of further special cases, we refer to [10].

(b) Discrete  $L_p$ -approximation: Let  $X$  be the  $m$ -dimensional space  $\mathfrak{R}^m$  with norms

$$\|y\|_p = \left( \sum_{i=1}^m |y_i|^p \right)^{1/p}$$

for  $1 \leq p < \infty$ , and

$$\|y\|_\infty = \max_{i=1, \dots, m} |y_i|,$$

where  $y = (y_1, \dots, y_m) \in \mathfrak{R}^m$ .

$X^*$  can be identified with  $\mathfrak{R}^m$ , and each element  $L \in X^*$  is given by

$$L(y) = \langle L, y \rangle = \sum_{i=1}^m l_i y_i, \quad (4.9)$$

where  $l = (l_1, \dots, l_m) \in \mathfrak{R}^m$  is uniquely defined by  $L$ . The norm of  $X^*$  is given by

$$\|L\| = \sup_{\|y\|_q=1} |L(y)| = \|l\|_q,$$

where  $(1/p) + (1/q) = 1$  for  $1 < p < \infty$ ,  $q = 1$  if  $p = \infty$ , and  $q = \infty$  if  $p = 1$ .

In order to apply Lemma 2.2, we consider a nonempty subset  $W \subseteq \mathfrak{R}^m$ , an element  $x \in \mathfrak{R}^m$ , and a nonempty subset  $\mathcal{L}$  of

$$S_q = \{l \in \mathfrak{R}^m : \|l\|_q = 1\}.$$

To each  $l \in \mathcal{L}$  we assign a real number  $\alpha_l$ , and define  $L \in X^*$  by (4.9). Then  $L \in S^*$ , and (4.1) is equivalent to the following statement: For each  $w \in W$ , there is an  $l \in \mathcal{L}$  such that

$$\langle l, w \rangle \geq \alpha_l. \tag{4.10}$$

By (4.2), we then have

$$\inf_{l \in \mathcal{L}} \{\alpha_l - \langle l, x \rangle\} \leq d_x^p(W) = \inf_{w \in W} \|w - x\|_p.$$

A very simple way of realizing (4.10) is the following: Let  $\mathcal{L}$  consist of  $m$  vectors of the form  $l^i = \epsilon_i e^i$ , where  $e^i = (e_1^i, \dots, e_m^i)$ ,  $e_j^i = \delta_{ij}$ , and  $\epsilon_i = +1$  or  $-1$ . Then, obviously,  $\|l^i\|_q = 1$  for every  $q$ ,  $1 \leq q \leq \infty$ . Putting  $\alpha_i = \alpha_{l^i}$ , we can express condition (4.10) by

$$\min_{i=1, \dots, m} \{\alpha_i - \epsilon_i w_i\} \leq 0 \tag{4.11}$$

for all  $w = (w_1, \dots, w_m) \in W$ .

Finally, we demonstrate in the case of rational approximation, how (4.11) can be realized:

Let  $U$  and  $V$  be subspaces of  $X = \mathfrak{R}^m$ , spanned by  $u^0, \dots, u^r$  and  $v^0, \dots, v^s$ , respectively, where  $r + s + 2 \leq m$ . We assume

$$V^+ = \{v \in V : v^i > 0, i = 1, \dots, m\}$$

to be nonempty, and put

$$W = \left\{ \frac{u}{v} : u \in U, v \in V^+ \right\}.$$

(4.11) is equivalent to the following statement: There is no vector  $(a_0, \dots, a_r, b_0, \dots, b_s) \in \mathfrak{R}^{r+s+2}$  such that

$$\left. \begin{aligned} \sum_{k=0}^s \alpha_i v_i^k b_k - \sum_{j=0}^r \epsilon_i u_i^j a_j &> 0 \\ \sum_{k=0}^s v_i^k b_k &> 0 \end{aligned} \right\} \text{for } i = 1, \dots, m.$$

By Theorem 2.9 of [4], this is equivalent to the existence of a vector  $(y_1, \dots, y_m, p_1, \dots, p_m) \in \mathfrak{R}^{2m}$  such that

$$\left. \begin{aligned} y_i &\geq 0 & \text{and} & & p_i &\geq 0, & i = 1, \dots, m, \\ \sum_{i=1}^m v_i^k \alpha_i y_i + \sum_{i=1}^m v_i^k p_i &= 0, & k = 0, \dots, s, \\ \sum_{i=1}^m u_i^j \epsilon_i y_i &= 0, & j = 0, \dots, r. \end{aligned} \right\} \quad (4.12)$$

We put  $c_i = -\epsilon_i y_i$  and  $\lambda_i = \alpha_i - \epsilon_i x_i$ , for  $i = 1, \dots, m$ . Then (4.11) is equivalent to the existence of vectors  $(c_1, \dots, c_m)$  and  $(p_1, \dots, p_m)$  of  $\mathfrak{R}^m$  such that

$$\left. \begin{aligned} p_i &\geq 0 & \text{for } i = 1, \dots, m, \\ \sum_{i=1}^m u_i^j c_i &= 0, & j = 0, \dots, r, \\ \sum_{i=1}^m v_i^k x_i c_i &= \sum_{i=1}^m v_i^k (\lambda_i |c_i| + p_i), & k = 0, \dots, s. \end{aligned} \right\} \quad (4.13)$$

Since  $V^+$  is assumed to be nonempty, it can easily be shown that not all  $c_i$  can vanish. If (4.13) is satisfied, we have by the definition of the  $\lambda_i$ 's,

$$\min_{i=1, \dots, m} \lambda_i \leq d_x^p(W).$$

Under the natural assumption that the matrix

$$\begin{pmatrix} u_i^j \\ v_i^k x_i \end{pmatrix}$$

has the rank  $r + s + 2$ , it is easy to satisfy (4.13). One merely has to choose  $z_i \geq 0$ ,  $i = 1, \dots, m$ , and compute a nontrivial solution  $(c_1, \dots, c_m)$  of

$$\begin{aligned} \sum_{i=1}^m u_i^j c_i &= 0, & j = 0, \dots, r, \\ \sum_{i=1}^m v_i^k x_i c_i &= \sum_{i=1}^m v_i^k z_i, & k = 0, \dots, s, \end{aligned}$$

which is always possible if not all the  $z_i$  vanish. We define

$$\lambda_i = \begin{cases} \frac{z_i}{|c_i|} & \text{if } c_i \neq 0, \\ \max_{c_i \neq 0} \frac{z_i}{|c_i|} & \text{if } c_i = 0, \end{cases}$$

and

$$p_i = \begin{cases} 0 & \text{if } c_i \neq 0, \\ z_i & \text{if } c_i = 0. \end{cases}$$

Then (4.13) is satisfied, and we have

$$\min_{c_i \neq 0} \frac{z_i}{|c_i|} \leq d_x^v(W).$$

#### REFERENCES

1. B. BROSOWSKI, Einige Bemerkungen zum verallgemeinerten Kolmogoroffschen Kriterium. Max-Planck-Institut für Physik und Astrophysik. MPI-PEA/Astro 7/68.
2. L. COLLATZ, Inclusion theorems for the minimal distance in rational Tschebyscheff approximation with several variables. In: "Approximation of Functions," (H. L. Garabedian, Ed.). Elsevier, Amsterdam-London-New York, 1965, 43-56.
3. F. R. DEUTSCH AND P. H. MASERICK, Applications of the Hahn-Banach theorem in approximation theory. *SIAM Review* **9** (1967), 516-530.
4. D. GALE, "Linear Economic Models." McGraw-Hill Book Company, New York-Toronto-London, 1960.
5. A. L. GARKAVI, Duality theorems for approximation by elements of convex sets (Russian). *Uspehi Mat. Nauk* **16** (1961), 141-145.
6. A. L. GARKAVI, On a criterion for an element of best approximation (Russian). *Sibirsk Mat. Z.* **5** (1964), 472-476.
7. S. J. HAVINSON, Approximation by elements of convex sets. *Soviet Math. Dokl.* **8** (1967), 98-101.
8. D. HENZE, Über die Menge der Minimallösungen bei linearen und nichtlinearen Approximationsproblemen. Schriften des Rheinisch-Westfälischen Instituts für instrumentelle Mathematik an der Universität Bonn (Herausgeber: E. Peschl, H. Unger), Serie A, Nr. 17, Köln-Opladen: Westdeutscher Verlag.
9. G. KÖTHE, "Topologische Lineare Räume I." Springer, Berlin-Göttingen-Heidelberg, 1960.
10. W. KRABS, Über differenzierbare asymptotisch konvexe Funktionenfamilien bei der nicht-linearen gleichmäßigen Approximation. *Arch. Rat. Mech. Anal.* **27** (1967), 275-288.
11. W. KRABS, Asymptotische Konvexität bei der Approximation in normierten Vektorräumen. *Math. Z.* **108** (1969), 368-376.
12. G. MEINARDUS AND D. SCHWEDT, Nichtlineare Approximationen. *Arch. Rat. Mech. Anal.* **17** (1964), 297-326.
13. V. N. NIKOLSKIĬ, Best approximation by elements of convex sets in normed linear spaces (Russian). *Kalinin. Gos. Ped. Inst. Ucen. Pap.* **29** (1963), 85-119.