

BEST SIMULTANEOUS APPROXIMATIONS FROM A CONVEX SUBSET

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1. Introduction

Let U and V be nonempty compact subsets of two Hausdorff topological vector spaces. Suppose that a function $J : U \times V \rightarrow \mathbb{R}$ is such that for each $v \in V$, $J(\cdot, v)$ is lower semi-continuous and convex on U , and for each $u \in U$, $J(u, \cdot)$ is upper semi-continuous and concave on V . Then, as is well known [2], there exists a saddle point $(u^*, v^*) \in U \times V$ such that

$$(1.1) \quad J(u^*, v) \leq J(u^*, v^*) \leq J(u, v^*), \quad u \in U, v \in V,$$

that is,

$$\min_{u \in U} \max_{v \in V} J(u, v) = \max_{v \in V} \min_{u \in U} J(u, v).$$

However, if V is not convex, or if for some $u \in U$, $J(u, \cdot)$ is not a concave function on V , then the relation (1.1) does not hold in general.

THEOREM 1.1.[4]. *Let U be an n -dimensional, compact convex set in a Hausdorff topological vector space ($n \geq 1$) and let V be a compact Hausdorff space. Let $J : U \times V \rightarrow \mathbb{R}$ be a jointly continuous function. Then $u^* \in U$ minimizes $\max_{v \in V} J(u, v)$ over U if and only if there exists $(\bar{\lambda}_{n+1}^*, \bar{v}_{n+1}^*) \in \bar{V}_{n+1}$ such that*

$$(1.2) \quad \sum_{i=1}^{n+1} \lambda_i J(u^*, v_i) \leq \sum_{i=1}^{n+1} \lambda_i^* J(u^*, v_i^*) \leq \sum_{i=1}^{n+1} \lambda_i^* J(u, v_i^*)$$

Received October 4, 1994.

1991 AMS Subject Classification: 41A28.

Key words and phrases: Best uniform approximation, Best simultaneous approximation.

*Supported partially by the Basic Science Research Institute program, Ministry of Education.

holds for all $(\bar{\lambda}_{n+1}, \bar{v}_{n+1}) \in \bar{V}_{n+1}$ and for all $u \in U$, where

$$\bar{V}_{n+1} = \{(\bar{\lambda}_n, \bar{v}_n) | \bar{\lambda}_n = (\lambda_1, \dots, \lambda_n), \bar{v} = (v_1, \dots, v_n), \\ \sum_{i=1}^n \lambda_i = 1, \lambda_i \geq 0, v_i \in V(i = 1, \dots, n)\}.$$

2. Best Uniform Approximations

Let X be a compact Hausdorff space and let Y be a normed linear space with a norm $\|\cdot\|$. Let $C(X, Y)$ denote the set of all continuous functions from X to Y . Let C be a convex set in an n -dimensional subspace of $C(X, Y)$ and $F \in C(X, Y)$. An element $f^* \in C$ is called a best uniform approximation to F from C if f^* minimizes

$$\max_{x \in X} \|f(x) - F(x)\|$$

over C , that is, if

$$\max_{x \in X} \|f^*(x) - F(x)\| \leq \max_{x \in X} \|f(x) - F(x)\|$$

holds for all $f \in C$. For $f \in C(X, Y)$, we define the uniform norm of f by

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

and endow the linear space $C(X, Y)$ with the uniform topology. Clearly $\|f(x) - F(x)\|$ is a jointly continuous function of two variables f, x and convex in f , i.e.,

$$\|(\lambda f + (1 - \lambda)g)(x) - F(x)\| \leq \lambda \|f(x) - F(x)\| + (1 - \lambda) \|g(x) - F(x)\|$$

for all $f, g \in C(X, Y)$, $x \in X$, and $\lambda, 0 \leq \lambda \leq 1$. Hence if C is a convex set in a finite dimensional subspace of $C(X, Y)$, we can apply Theorem 1.1 and obtain the following necessary and sufficient condition of a best uniform approximation from C .

THEOREM 2.1. *Let C be a closed convex set in an n -dimensional subspace of $C(X, Y)$ and $F \in C(X, Y)$. Then $f^* \in C$ is a best uniform approximation to F from C if and only if there exist $\lambda_1^*, \dots, \lambda_k^* > 0$, $\sum_{i=1}^k \lambda_i^* = 1$, and k distinct elements x_1^*, \dots, x_k^* of X , where $1 \leq k \leq n + 1$, satisfying*

$$(i) \quad \|f^*(x_i^*) - F(x_i^*)\| = \|f^* - F\| \quad i = 1, 2, \dots, k:$$

$$(ii) \quad \sum_{i=1}^k \lambda_i^* \|f(x_i^*) - F(x_i^*)\| \geq \sum_{i=1}^k \lambda_i^* \|f^*(x_i^*) - F(x_i^*)\|$$

for any $f \in C$.

Proof. (\Rightarrow) Let f^* be a best uniform approximation to F from C and let $U = \{f \in C \mid \|f^* - f\| \leq 1\}$. Then f^* minimizes $\max_{x \in X} \|f(x) - F(x)\|$ over U and U is a compact subset of C because it is bounded and closed in an n -dimensional subspace. By Theorem 1.1, there exist $\lambda'_1, \dots, \lambda'_{n+1} \geq 0$, $\sum_{i=1}^{n+1} \lambda'_i = 1$, and $\{x'_1, \dots, x'_{n+1}\} \subset X$ such that

$$(2.1) \quad \sum_{i=1}^{n+1} \lambda'_i \|f^*(x'_i) - F(x'_i)\| \geq \sum_{i=1}^{n+1} \lambda_i \|f^*(x_i) - F(x_i)\|$$

for all $(\bar{\lambda}_{n+1}, \bar{x}_{n+1}) \in \bar{X}_{n+1} = \{(\bar{\lambda}_{n+1}, \bar{x}_{n+1}) : \bar{\lambda}_{n+1} = (\lambda_1, \dots, \lambda_{n+1}),$

$$\bar{x}_{n+1} = (x_1, \dots, x_{n+1}), \sum_{i=1}^{n+1} \lambda_i = 1, \lambda_i \geq 0, x_i \in X (i = 1, \dots, n + 1)\};$$

$$(2.2) \quad \sum_{i=1}^{n+1} \lambda'_i \|f(x'_i) - F(x'_i)\| \geq \sum_{i=1}^{n+1} \lambda'_i \|f^*(x'_i) - F(x'_i)\|$$

for all $f \in U$. Let us denote the nonzero elements within $\lambda'_1, \dots, \lambda'_{n+1}$ by $\lambda_1^*, \dots, \lambda_k^*$ and the corresponding elements within x'_1, \dots, x'_{n+1} of X by x_1^*, \dots, x_k^* . The assertion (i) follows from (2.1) which means, for $i = 1, \dots, k$,

$$\|f^*(x_i^*) - F(x_i^*)\| = \max_{x \in X} \|f^*(x) - F(x)\| = \|f^* - F\|.$$

On the other hand, it follows from (2.2) that

$$\sum_{i=1}^k \lambda_i^* \|f(x_i^*) - F(x_i^*)\| \geq \sum_{i=1}^k \lambda_i^* \|f^*(x_i^*) - F(x_i^*)\|$$

holds for all $f \in U$. Since the left-hand side of the above inequality is a convex function of f and has a local minimum at f^* , f^* realizes a global minimum by a property of convex functions. Thus ii) holds.

(\Leftarrow) Conversely, suppose that (i) and (ii) hold. These two conditions yield

$$\begin{aligned} \sup_{(\bar{\lambda}_k, \bar{x}_k) \in \bar{X}_k} \inf_{f \in C} \sum_{i=1}^k \lambda_i \|f(x_i) - F(x_i)\| &\geq \sum_{i=1}^k \lambda_i^* \|f^*(x_i^*) - F(x_i^*)\| \\ &= \| \|f^* - F\| \|, \end{aligned}$$

where \bar{X}_k is similarly defined as the set \bar{X}_{n+1} . The left-hand side of the last relation is equal to or less than

$$\inf_{f \in C} \max_{(\bar{\lambda}_k, \bar{x}_k) \in \bar{X}_k} \sum_{i=1}^k \lambda_i \|f(x_i) - F(x_i)\| = \inf_{f \in C} \max_{x \in X} \|f(x) - F(x)\|,$$

so

$$\| \|f^* - F\| \| \leq \inf_{f \in C} \| \|f - F\| \|.$$

Therefore, f^* is a best uniform approximation to F from C .

REMARK. The proof of Theorem 2.1 is the same as that of Theorem 4.1 in [4].

The next corollary states that f^* is a best uniform approximation on X if and only if it also is on some finite subset of X .

COROLLARY 2.2. *Let C be a closed convex set in an n -dimensional subspace of $C(X, Y)$ and $F \in C(X, Y)$. Then f^* is a best uniform approximation to F from C if and only if there exist k elements $x_1^*, \dots, x_k^* \in X$, where $1 \leq k \leq n + 1$, for which f^* satisfies*

- (i) $\|f^*(x_j^*) - F(x_j^*)\| = \| \|f^* - F\| \|, \quad j = 1, \dots, k;$
- (ii) $\max_{1 \leq j \leq k} \|f^*(x_j^*) - F(x_j^*)\| \geq \max_{1 \leq j \leq k} \|f^*(x_j^*) - F(x_j^*)\|$

for all $f \in C$.

COROLLARY 2.3 [3]. *Let C be an n -dimensional subspace of $C(X, Y)$ and $F \in C(X, Y)$. Then $f^* \in C$ is a best uniform approximation to F from C if and only if there exist $\lambda_1^*, \dots, \lambda_k^* > 0$, $\sum_{i=1}^k \lambda_i^* = 1$, and k distinct elements x_1^*, \dots, x_k^* of X , where $1 \leq k \leq n + 1$, satisfying*

$$(i) \quad \|f^*(x_i^*) - F(x_i^*)\| = \|f^* - F\|, \quad i = 1, \dots, k;$$

$$(ii) \quad \sum_{i=1}^k \lambda_i^* \|f^*(x_i^*) - F(x_i^*)\| \geq \sum_{i=1}^k \lambda_i^* \|f^*(x_i^*) - F(x_i^*)\|$$

for all $f \in C$.

Proof. The conclusion holds since an n -dimensional subspace is a closed convex set.

3. Best Simultaneous Approximations

Suppose that functions F_1, \dots, F_ℓ are in $C(X, Y)$. Define a norm on the linear space of ℓ -tuples of elements of $C(X, Y)$ as follows: for any F_1, \dots, F_ℓ in $C(X, Y)$ let $\mathbf{F} = (F_1, \dots, F_\ell)$ and

$$\|\|\mathbf{F}\|\| = \| (F_1, \dots, F_\ell) \| = \max_{\|\mathbf{a}\|_1} \left\| \sum_{i=1}^{\ell} a_i F_i \right\|.$$

We want to approximate these functions simultaneously by functions in a closed convex set C in an n -dimensional subspace of $C(X, Y)$. That is, the problem is to find a function $f \in C$ which minimizes

$$(3.1) \quad \max_{1 \leq j \leq \ell} \max_{x \in X} \|F_j(x) - f(x)\|$$

over C . If such a function f^* in C exists, we call it a best simultaneous approximation for $\mathbf{F} = (F_1, \dots, F_\ell)$ from C .

In this section, we want to derive a necessary and sufficient condition for a function to be a best simultaneous approximation. It is easy to show that (3.1) can be expressed as

$$\max_{\mathbf{a} \in A} \max_{x \in X} \left\| \sum_{j=1}^{\ell} \alpha_j F_j(x) - f(x) \right\| = \max_{\mathbf{a} \in A} \max_{x \in X} \|\mathbf{a}\mathbf{F}(x) - f(x)\|$$

where the set A is defined by

$$A = \{\mathbf{a} = (\alpha_1, \dots, \alpha_\ell) : \sum_{j=1}^{\ell} \alpha_j = 1, \alpha_j \geq 0 (1 \leq j \leq \ell)\}.$$

If $\sum_{j=1}^{\ell} \alpha_j F_j(x)$ is denoted by the inner product $\mathbf{a}\mathbf{F}(x)$ of two vectors $\mathbf{a} = (\alpha_1, \alpha_2, \dots, \alpha_\ell)$ and $\mathbf{F}(x) = (F_1(x), \dots, F_\ell(x))$, then the problem takes on the expression

$$(3.2) \quad \text{minimize} \quad \max_{(\mathbf{a}, x) \in A \times X} \|\mathbf{a}\mathbf{F}(x) - f(x)\| \text{ over the set } C.$$

Note that $A \times X$ is compact, and $\|\mathbf{a}\mathbf{F}(x) - f(x)\|$ is a jointly continuous function of three variables \mathbf{a} , x , f and convex in f , that is,

$$\|\mathbf{a}\mathbf{F}(x) - (\theta f + (1-\theta)g)(x)\| \leq \theta \|\mathbf{a}\mathbf{F}(x) - f(x)\| + (1-\theta) \|\mathbf{a}\mathbf{F}(x) - g(x)\|$$

for all $f, g \in C$ and $\theta, 0 \leq \theta \leq 1$.

THEOREM 3.1. *Let C be a closed convex set in an n -dimensional subspace of $C(X, Y)$ and let $\mathbf{F} = (F_1, \dots, F_\ell)$. Then $f^* \in C$ is a best simultaneous approximation for \mathbf{F} from C if and only if there exist $\lambda_1^*, \dots, \lambda_k^* > 0$, $\sum_{i=1}^k \lambda_i^* = 1$, k distinct elements $x_1^*, \dots, x_k^* \in X$, and k vectors $\mathbf{a}_1^*, \dots, \mathbf{a}_k^* \in A$, where $1 \leq k \leq n + 1$, such that*

$$(i) \quad \begin{aligned} \|\mathbf{a}_i^* \mathbf{F}(x_i^*) - f^*(x_i^*)\| &= \max_{1 \leq j \leq \ell} \|F_j(x_i^*) - f^*(x_i^*)\| \\ &= \max_{1 \leq j \leq \ell} \|F_j - f^*\| \quad i = 1, \dots, k; \end{aligned}$$

(ii)

$$\sum_{i=1}^k \lambda_i^* \|\mathbf{a}_i^* \mathbf{F}(x_i^*) - f^*(x_i^*)\| \geq \sum_{i=1}^k \lambda_i^* \|\mathbf{a}_i^* \mathbf{F}(x_i^*) - f^*(x_i^*)\|$$

for any $f \in C$.

Proof. By the preceding remarks concerning the function $\|\mathbf{a}\mathbf{F}(x) - f(x)\|$, we can apply Theorem 1.1. The rest of the proof is similar to that of Theorem 2.1.

Recall that the subdifferential or set of subgradients of $|||\cdot|||$ at any element f of $C(X, Y)$ is the set defined by

$$\partial|||f||| = \{w \in C^*(X, Y) : |||g||| \geq |||f||| \\ + \langle w, g - f \rangle \text{ for any } g \in C(X, Y)\}$$

where the usual inner product notation is used to link elements of $C(X, Y)$ and its dual $C^*(X, Y)$. For an ℓ -tuple $\Phi = (\phi_1, \dots, \phi_\ell)$ of functions in $C(X, Y)$, define the set

$$G(\Phi) = \{(\mathbf{a}, w) : \mathbf{a} \in \mathbb{R}^\ell, |||\mathbf{a}|||_1 = 1, \\ \sum_{i=1}^{\ell} a_i \phi_i = |||\Phi|||u, |||u||| = 1, w \in \partial|||u|||\}.$$

Now let

$$\phi_i(f) = F_i - f, \quad i = 1, \dots, \ell$$

for all $f \in C$ and let $H(f)$ denote the set of ℓ -tuples $\{h_1, \dots, h_\ell\}$ of elements in $C^*(X, Y)$ defined by

$$H(f) = \text{conv}\{(a_1\omega, \dots, a_\ell\omega) : (\mathbf{a}, \omega) \in G(\Phi(f))\},$$

where as usual 'conv' is used to denote the convex hull. Note that

$$\sum_{i=1}^{\ell} \langle h_i, \phi_i(f) \rangle = |||\Phi(f)||| \text{ for all } \mathbf{h} \in H(f).$$

For $f \in C$ and any $\mathbf{a} \in \mathbb{R}^\ell$ with $|||\mathbf{a}|||_1 = 1$, define

$$E(\mathbf{a}, f) = \{x \in X : |||\sum_{i=1}^{\ell} a_i \phi_i(f(x))|||_Y = |||\Phi(f)|||\},$$

and let

$$E(f) = \{x \in X : |||\sum_{i=1}^{\ell} a_i \phi_i(f(x))|||_Y = |||\Phi(f)||| \text{ for some } \mathbf{a}, |||\mathbf{a}||| = 1\}.$$

Then for each distinct $x_k \in E(f)$, there exists $\mathbf{a}_k, \|\mathbf{a}_k\|_1 = 1$, such that

$$(3.3) \quad x_k \in E(\mathbf{a}_k, f).$$

In particular, $(\mathbf{a}, w) \in G(\Phi(f))$ implies that

$$w \in \text{conv}\{v(x)\delta(x) : x \in E(\mathbf{a}, f)\},$$

where δ is the delta function, and $v(x) \in \partial\|\sum_{i=1}^{\ell} \alpha_i \phi_i(f(x))\|_Y$. Thus

$$\langle w, f \rangle = \sum_{k=1}^s \mu_k \langle v(x_k), f(x_k) \rangle_Y,$$

where $\langle \cdot, \cdot \rangle_Y$ denotes an inner product between elements of Y and its dual, where $x_k \in E(\mathbf{a}, f)$ and $\mu_k, k = 1, \dots, s$, are nonnegative numbers summing to 1.[See [5]]

COROLLARY 3.2. *Let C be a closed convex set in an n -dimensional subspace of $C(X, Y)$. Then $f^* \in C$ is a best simultaneous approximation for \mathbf{F} from C if and only if there exist $m(\leq n + 1)$ distinct elements x_1, \dots, x_m of $E(f^*)$, m vectors $\mathbf{a}_1, \dots, \mathbf{a}_m$ satisfying $x_k \in E(\mathbf{a}_k, f^*)(k = 1, \dots, m)$ and m positive numbers $\gamma_1, \dots, \gamma_m$ summing to 1 such that*

$$\sum_{k=1}^m \gamma_k \|\mathbf{a}_k \mathbf{F}(x_k) - f(x_k)\| \geq \sum_{k=1}^m \gamma_k \|\mathbf{a}_k \mathbf{F}(x_k) - f^*(x_k)\|$$

for any $f \in C$.

Proof. By Theorem 3.1 and the above argument, f^* is a best simultaneous approximation for \mathbf{F} from C if and only if there exist $\gamma_1, \dots, \gamma_m > 0, \sum_{k=1}^m \gamma_k = 1, m$ distinct elements $x_1, \dots, x_m \in X$ and k vectors $\mathbf{a}_1, \dots, \mathbf{a}_m \in A$, where $1 \leq k \leq n + 1$, such that

$$(1) \quad \begin{aligned} \|\mathbf{a}_k \mathbf{F}(x_k) - f^*(x_k)\| &= \max_{1 \leq j \leq \ell} \|F_j(x_k) - f^*(x_k)\| \\ &= \max_{1 \leq j \leq \ell} \|F_j - f^*\| \quad k = 1, \dots, m \end{aligned}$$

$$(2) \quad \sum_{k=1}^m \gamma_k \|\mathbf{a}_k \mathbf{F}(x_k) - f(x_k)\| \geq \sum_{k=1}^m \gamma_k \|\mathbf{a}_k \mathbf{F}(x_k) - f^*(x_k)\|$$

for any $f \in C$ if and only if there exist $m(\leq n + 1)$ distinct elements x_1, \dots, x_m of $E(f^*)$, m vectors $\mathbf{a}_1, \dots, \mathbf{a}_m$ satisfying $x_k \in E(\mathbf{a}_k, f^*)(k = 1, \dots, m)$ and m positive numbers $\gamma_1, \dots, \gamma_m$ summing to 1 such that

$$\sum_{k=1}^m \gamma_k \|\mathbf{a}_k \mathbf{F}(x_k) - f(x_k)\| \geq \sum_{k=1}^m \gamma_k \|\mathbf{a}_k \mathbf{F}(x_k) - f^*(x_k)\|$$

for any $f \in C$.

Since an n -dimensional subspace is a closed convex set, we obtain the next corollary.

COROLLARY 3.3. [4]. *Let C be an n -dimensional subspace of $C(X, Y)$. Then $f^* \in C$ is a best simultaneous approximation for \mathbf{F} from C if and only if there exist $\lambda_1^*, \dots, \lambda_k^* > 0$, $\sum_{i=1}^k \lambda_i^* = 1$, k distinct elements $x_1^*, \dots, x_k^* \in X$, and k vectors $\mathbf{a}_1^*, \dots, \mathbf{a}_k^* \in A$, where $1 \leq k \leq n + 1$, such that*

$$(i) \quad \begin{aligned} \|\mathbf{a}_i^* \mathbf{F}(x_i^*) - f^*(x_i^*)\| &= \max_{1 \leq j \leq \ell} \|F_j(x_i^*) - f^*(x_i^*)\| \\ &= \max_{1 \leq j \leq \ell} \|\|F_j - f^*\|\| \quad i = 1, \dots, k; \end{aligned}$$

(ii)

$$\sum_{i=1}^k \lambda_i^* \|\mathbf{a}_i^* \mathbf{F}(x_i^*) - f(x_i^*)\| \geq \sum_{i=1}^k \lambda_i^* \|\mathbf{a}_i^* \mathbf{F}(x_i^*) - f^*(x_i^*)\|$$

for any $f \in C$.

COROLLARY 3.4. *Let C be a closed convex subset in an n -dimensional subspace of $C(X, Y)$ and let $\mathbf{F} = (F_1, \dots, F_\ell)$. If there exist $m(\leq n + 1)$ distinct elements x_1, \dots, x_m of $E(f^*)$, m vectors $\mathbf{a}_1, \dots, \mathbf{a}_m$ satisfying with $f = f^*$, and m positive numbers $\gamma_1, \dots, \gamma_m$ summing to 1 such that*

$$(i) \quad \sum_{k=1}^m \gamma_k \langle v_k(x_k), f^*(x_k) \rangle_Y = 0;$$

$$(ii) \quad \sum_{k=1}^m \gamma_k \langle v_k(x_k), f(x_k) \rangle_Y \leq 0$$

for any $f \in C$, where $v_k(x) \in \partial \|\sum_{i=1}^{\ell} a_k^i \phi_i(f^*(x))\|_Y, k = 1, 2, \dots, m$, then $f^* \in C$ is a best simultaneous approximation for \mathbf{F} from C .

Proof. By Theorem 3.1, it suffices to show that

$$\sum_{k=1}^m \gamma_k \|\sum_{i=1}^{\ell} a_k^i \phi_i(f^*(x_k))\|_Y \leq \sum_{k=1}^m \gamma_k \|\sum_{i=1}^{\ell} a_k^i \phi_i(f(x_k))\|_Y$$

for any $f \in C$. Let $f \in C$ be arbitrary, and let

$$v_k(x) \in \partial \|\sum_{i=1}^{\ell} a_k^i \phi_i(f^*(x_k))\|_Y \quad k = 1, 2, \dots, m$$

satisfy (i) and (ii). Then

$$\begin{aligned} & \sum_{k=1}^m \gamma_k \|\sum_{i=1}^{\ell} a_k^i \phi_i(f^*(x_k))\|_Y \\ &= \sum_{k=1}^m \gamma_k \langle v_k(x_k), \sum_{i=1}^{\ell} a_k^i (F_i(x_k) - f^*(x_k)) \rangle_Y \\ &\leq \sum_{k=1}^k \gamma_k \langle v_k(x_k), \sum_{i=1}^{\ell} a_k^i (F_i(x_k) - f(x_k)) \rangle_Y \\ &\leq \sum_{k=1}^m \gamma_k \|\sum_{i=1}^{\ell} a_k^i \phi_i(f(x_k))\|_Y. \end{aligned}$$

Thus f^* is a best simultaneous approximation for \mathbf{F} from C .

COROLLARY 3.5. Let C be a convex subset in an n -dimensional subspace of $C(X, Y)$ and let $\mathbf{F} = (F_1, \dots, F_{\ell})$. If there exists $\mathbf{h} = (h_1, h_2, \dots, h_{\ell}) \in H(f^*)$ such that

$$(i) \quad \sum_{i=1}^{\ell} \langle h_i, f - f^* \rangle \leq 0$$

for all $f \in C$, then $f^* \in C$ is a best simultaneous approximation for \mathbf{F} from C .

Proof. Let f be any element of C . Then

$$\begin{aligned} \|\phi(f)\| &= \max_{\|\mathbf{a}\|_1=1} \left\| \sum_{i=1}^{\ell} a_i(F_i - f) \right\|_A \\ &\geq \langle w, \sum_{i=1}^{\ell} a_i(F_i - f) \rangle \end{aligned}$$

for all $(\mathbf{a}, w) \in G(\Phi(f^*))$. Suppose that $\mathbf{h} \in H(f^*)$ satisfies (i). Then

$$\begin{aligned} \|\phi(f)\| &\geq \sum_{i=1}^{\ell} \langle h_i, F_i - f \rangle \\ &= \sum_{i=1}^{\ell} \langle h_i, F_i - f^* + f^* - f \rangle \\ &\geq \sum_{i=1}^{\ell} \langle h_i, F_i - f^* \rangle \\ &= \|\phi(f^*)\|. \end{aligned}$$

Thus f^* is a best simultaneous approximation for \mathbf{F} from C .

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