A Characterization of Best Algebraic Approximations from Below and from Above in the Multivariate Case

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Abstract

In this paper the constrained K-functionals connected with the best multivariate algebraic approximations from below and from above are characterized in terms of moduli of smoothness. The results are a multivariate generalization of those in [2].

1 Introduction.

We consider measurable real-valued bounded (from below or from above) functions defined in every point of the domain $\Omega = \Pi[-1; 1]$, where

$$\Pi[\mathbf{a};\mathbf{b}] \stackrel{\text{def}}{=} \left\{ \mathbf{x} \in \mathbb{R}^d \mid x_i \in [\min\{a_i, b_i\}, \max\{a_i, b_i\}] \quad for \ every \ i = 1, \dots, d \right\}.$$

 \mathbb{R}^d is considered as a normed vector space with elements $\mathbf{x} = (x_1, \ldots, x_d)$, \mathbf{a} , \mathbf{b} , \mathbf{y} , \mathbf{h} and norm $\|\mathbf{x}\| = \max\{|x_1|, \ldots, |x_d|\}$. Here $\mathbf{1}$ and $-\mathbf{1}$ mean respectively $(1, \ldots, 1)$ and $(-1, \ldots, -1)$.

Let X be a measurable subset of Ω . We shall consider the following spaces

$$L_p(X) \stackrel{\text{def}}{=} \left\{ f \mid ||f||_{p(X)} = \left\{ \int_X |f(\mathbf{x})|^p d\mathbf{x} \right\}^{\frac{1}{p}} < \infty \right\},$$

for $p \in [1, \infty)$ (dx means the Lebesgue measure on X) and

$$L_{\infty}(X) \stackrel{\text{def}}{=} \left\{ f \mid \|f\|_{\infty(X)} = ess \, sup \left\{ |f(\mathbf{x})| \; ; \; \mathbf{x} \in X \right\} < \infty \right\},$$

for $p = \infty$.

 α, β are multi-indices. If $\alpha = (\alpha_1, ..., \alpha_d), \alpha_s \ge 0$ for any $s = 1, ..., d, |\alpha| = \sum_{i=1}^d \alpha_i$ is the length of α . $\alpha \ge \beta$ means $\alpha_s \ge \beta_s$ for any $s = 1, ..., d, \alpha! = \prod_{s=1}^d \alpha_s!$ and $\binom{\alpha}{\beta} = \prod_{s=1}^d \binom{\alpha_s}{\beta_s}$.

Let r be natural. By $W_p^r(X)$ we denote the Sobolev space

$$W_p^r(X) \stackrel{\text{def}}{=} \left\{ f \mid \sum_{|\alpha|=r} \|D^{\alpha}f\|_{p(X)} < \infty \right\}, \quad where \quad D^{\alpha} = \prod_{i=1}^d \frac{\partial^{\alpha_i}}{\partial x_i^{\alpha_i}}$$

For $v \in [-1,1]$, t > 0 we set $\psi(t,v) \stackrel{\text{def}}{=} t\sqrt{1-v^2} + t^2$. For $\mathbf{x} \in \Omega$ we denote $\Psi(t,\mathbf{x}) \stackrel{\text{def}}{=} \prod_{s=1}^d \psi(t,x_s)$ and $\Psi^{\alpha}(t,\mathbf{x}) \stackrel{\text{def}}{=} \prod_{s=1}^d \psi(t,x_s)^{\alpha_s}$. A t neighbourhood of the point $\mathbf{x} \in \Omega$ we define by

 $U(t, \mathbf{x}) \stackrel{\text{def}}{=} \{ \mathbf{y} \in \Omega \mid |x_s - y_s| \le \psi(t, x_s) \text{ for every } s = 1, ..., d \}.$

Everywhere in this paper c denotes a positive number which may depend on r and d. The c's may differ at each occurrence. If c depends on another parameter we indicate this using indices.

By H_n we denote the set of all algebraic polynomials in \mathbb{R}^d of total degree not greater than n. The best approximations by algebraic polynomials are given by

$$E(f, H_n)_{p(X)} \stackrel{\text{def}}{=} \inf \left\{ \|f - Q\|_{p(X)} \mid Q \in H_n \right\}$$

and the best approximations from below or from above by algebraic polynomials are given respectively by

(1.1)
$$E^{-}(f, H_{n})_{p(X)} \stackrel{\text{def}}{=} \inf \left\{ \|f - Q\|_{p(X)} \mid Q \in H_{n} , Q \leq f \right\}$$

and

(1.2)
$$E^+(f, H_n)_{p(X)} \stackrel{\text{def}}{=} \inf \left\{ \|f - Q\|_{p(X)} \mid Q \in H_n , Q \ge f \right\},$$

whenever f is bounded from below or from above respectively.

Let $l = \max\left\{\left[\frac{d}{p}\right] + 1, r\right\}$ ([·] – integral part). We investigate the K-functionals

(1.3)
$$K_r^-(f,t)_p = K^-\left(f,\Psi(t);L_p,W_p^r,W_p^l\right)$$

$$\stackrel{\text{def}}{=} \inf\left\{ \|f-g\|_{p(\Omega)} + \sum_{|\alpha|=r,l} \|\Psi^{\alpha}(t)D^{\alpha}g\|_{p(\Omega)} \mid g \le f , g \in W_p^l(\Omega) \right\},$$

(1.4)
$$K_{r}^{+}(f,t)_{p} = K^{+}\left(f,\Psi(t);L_{p},W_{p}^{r},W_{p}^{l}\right)$$

$$\stackrel{\text{def}}{=} \inf\left\{\|f-g\|_{p(\Omega)}+\sum_{|\alpha|=r,l}\|\Psi^{\alpha}(t)D^{\alpha}g\|_{p(\Omega)} \mid g \geq f , g \in W_{p}^{l}(\Omega)\right\}$$

and

$$K\left(f,\Psi(t);L_p,W_p^r\right) \stackrel{\text{def}}{=} \inf\left\{ \|f-g\|_{p(\Omega)} + \sum_{|\alpha|=r} \|\Psi^{\alpha}(t)D^{\alpha}g\|_{p(\Omega)} \mid g \in W_p^r(\Omega) \right\}.$$

In [5] we prove the following direct and inverse inequalities for the best constrained approximations in terms of the K-functionals. **Theorem 1.1** Let $1 \le p \le \infty$, let r and n be natural, * = "-" or "+" and let $f \in L_p(\Omega)$ be bounded from below or from above respectively. Then we have

(d)
$$E^*(f, H_{n-1})_{p(\Omega)} \le cK^*\left(f, \Psi(n^{-1}); L_p, W_p^r, W_p^l\right);$$

(i) $U^*(f, \Psi(n^{-1}), L_p, W_p^r, W_p^l) = (CF^*(f, W_p^r), W_p^l)$

(i)
$$K^*(f, \Psi(n^{-1}); L_p, W_p^i, W_p^i) \le c(E^*(f, H_{n-1})_{p(\Omega)} + K(f, \Psi(n^{-1}); L_p, W_p^i))$$

This inequalities are the reason for the investigation in this paper.

The main result of this paper Theorem 1.5 is a characterization for r = 1 and r = 2 of the K-functional (1.3) in terms of appropriate moduli. As a corollary we give a characterization of the best algebraic approximations from below. Similar results for the K-functional (1.4) and for the best algebraic approximations from above follow as a corollary from $E^+(f) = E^-(-f)$, $K^+(f) = K^-(-f)$ (with one and the same values of the parameters).

The equivalence between the K-functional from below and a characteristic based on local approximations from below by algebraic polynomials we give in

Theorem 1.2 Let $f \in L_p(\Omega)$ $(p \in [1, \infty])$ be bounded from below and let r be a natural number. Then

$$K_r^-(f,t)_p \sim \|\Psi(t,\cdot)^{-\frac{1}{p}} E^-(f,H_{r-1})_{p(U(t,\cdot))}\|_{p(\Omega)} \text{ for } t \in (0,1].$$

Remark 1.1. We consider $K_r^-(f,t)_p$ with argument $t \in (0,1]$ because of Theorem 1.1. Let $U \subset \mathbb{R}^d$ be a convex body. We set

(1.5)
$$\omega_r(f,U)_p \stackrel{\text{def}}{=} \sup\left\{ \|\Delta_{\mathbf{h},U}^r f(\cdot)\|_{p(U)} \mid \mathbf{h} \in \mathbb{R}^d \right\}$$

where

$$\Delta_{\mathbf{h},U}^{r} f(\mathbf{x}) \stackrel{\text{def}}{=} \begin{cases} \Delta_{\mathbf{h}}^{r} f(\mathbf{x}) & if \ \mathbf{x}, \mathbf{x} + r\mathbf{h} \in U; \\ 0 & otherwise \end{cases}$$

and

$$\Delta_{\mathbf{h}}^{r} f(\mathbf{x}) \stackrel{\text{def}}{=} \sum_{i=0}^{r} (-1)^{r-i} \binom{r}{i} f(\mathbf{x}+i\mathbf{h})$$

In order to handle the cases of approximations from below for r = 1 and r = 2 we introduce the following characteristics

(1.6)
$$\tau_r^-(f,U)_p \stackrel{\text{def}}{=} \|\sup\{\tilde{\Delta}_{\mathbf{h},U}^r f(\cdot) \mid \mathbf{h} \in \mathbb{R}^d\}\|_{p(U)}$$

and

$$\tau_r^-(f,\Psi(t))_{p(\Omega)} \stackrel{\text{def}}{=} \|\sup\{\tilde{\Delta}^r_{\mathbf{h},U(t,\cdot)}f(\cdot) \mid \mathbf{h} \in \mathbb{R}^d\}\|_{p(\Omega)},$$

where

$$\tilde{\Delta}^{1}_{\mathbf{h},U}f(\mathbf{x}) \stackrel{\text{def}}{=} \begin{cases} f(\mathbf{x}) - f(\mathbf{x} + \mathbf{h}) & if \ \mathbf{x}, \ \mathbf{x} + \mathbf{h} \in U; \\ 0 & otherwise \end{cases}$$

and

$$\tilde{\Delta}_{\mathbf{h},U}^2 f(\mathbf{x}) \stackrel{\text{def}}{=} \begin{cases} 2f(\mathbf{x}) - f(\mathbf{x} + \mathbf{h}) - f(\mathbf{x} - \mathbf{h}) & if \ \mathbf{x} - \mathbf{h}, \ \mathbf{x} + \mathbf{h} \in U; \\ 0 & otherwise. \end{cases}$$

We use the above characteristics in the following Whitney-type Theorem

Theorem 1.3 Let $f \in L_p(\Pi)$ $(\Pi = \Pi[\mathbf{a}; \mathbf{b}], p \in [1, \infty])$ is bounded from below. Then

(I)
$$E^{-}(f, H_0)_{p(\Pi)} = \tau_1^{-}(f, \Pi)_p;$$

(II) $E^{-}(f, H_1)_{p(\Pi)} \sim \omega_2(f, \Pi)_p + \tau_2^{-}(f, \Pi)_p$

Theorem 1.3 is proved in the Section 3.

Let $\Pi = \Pi[\mathbf{a}; \mathbf{b}]$ and $\pi = \Pi[\mathbf{c}; \mathbf{d}]$ be such that

(1.7)
$$\pi \subseteq \left(\Pi - \frac{\mathbf{a} + \mathbf{b}}{2}\right) \subseteq R.\pi$$

for some $R\geq 1,$ where for $U\subset \mathbb{R}^d$, $\mathbf{y}\in \mathbb{R}^d$ and t>0 we denote

$$U + \mathbf{y} \stackrel{\text{def}}{=} \{ \mathbf{x} \in \mathbb{R}^d \mid \mathbf{x} - \mathbf{y} \in U \}$$

and

$$tU \stackrel{\text{def}}{=} \{ \mathbf{x} \in \mathbb{R}^d \mid t^{-1}\mathbf{x} \in U \}.$$

We use the following characteristic of f.

(1.8)
$$\tau_r(f,\pi)_{p,p(\Pi)} \stackrel{\text{def}}{=} \left\{ \int_{\Pi} \frac{1}{\mu(\pi)} \int_{\pi} |\Delta^r_{\mathbf{v},\Pi} f(\mathbf{x})|^p d\mathbf{v} d\mathbf{x} \right\}^{\frac{1}{p}}.$$

Here $\mu(V)$ denotes the Lebesgue measure of the measurable set V. A relationship between (1.5) and (1.8) is established in [4, Sec.3]. The statement is

Theorem 1.4 If (1.7) is satisfied and $f \in L_p(\Omega)$ $(p \in [1, \infty])$ then,

$$c\tau_r(f,\pi)_{p,p(\Pi)} \le \omega_r(f,\Pi)_p \le cR^{d+r}\tau_r(f,\pi)_{p,p(\Pi)}.$$

We set

$$B(t, \mathbf{x}) \stackrel{\text{def}}{=} \left\{ \mathbf{y} \in \mathbb{R}^d \mid |y_s| \le \psi(t, x_s) \text{ for every } s = 1, ..., d \right\}.$$

In this paper we investigate the following averaged modulus of smoothness

(1.9)
$$\tau_r(f,\Psi(t))_{p,p(\Omega)} \stackrel{\text{def}}{=} \left\{ \int_{\Omega} \Psi(t,\mathbf{x})^{-1} \int_{B(t,\mathbf{x})} |\Delta_{\mathbf{v},\Omega}^r f(\mathbf{x})|^p d\mathbf{v} d\mathbf{x} \right\}^{\frac{1}{p}}.$$

Using the results from Sections 2, 3, Theorem 1.1 and Theorem 1.4 in Section 4 we give a characterization of the constrained K-functional in terms of appropriate moduli.

Theorem 1.5

$$\begin{split} K^{-}(f,\Psi(t),L_{p},W_{p}^{1},W_{p}^{l_{1}}) &\sim \tau_{1}^{-}(f,\Psi(t))_{p(\Omega)}, \qquad l_{1} = \left[\frac{d}{p}\right] + 1; \\ K^{-}(f,\Psi(t),L_{p},W_{p}^{2},W_{p}^{l_{2}}) &\sim \tau_{2}^{-}(f,\Psi(t))_{p(\Omega)} + \tau_{2}(f,\Psi(t))_{p,p(\Omega)}, \quad l_{2} = \max\{2,\left[\frac{d}{p}\right] + 1\} \end{split}$$

Combining the results of Theorem 1.1 and Theorem 1.5 in Section 4 we give a characterization of best approximation from below in terms of appropriate moduli.

Theorem 1.6

$$E^{-}(f, H_{n-1})_{p(\Omega)} \leq c\tau_{1}^{-}(f, \Psi(n^{-1}))_{p(\Omega)};$$

$$E^{-}(f, H_{n-1})_{p(\Omega)} \leq c\{\tau_{2}^{-}(f, \Psi(n^{-1}))_{p(\Omega)} + \tau_{2}(f, \Psi(n^{-1}))_{p,p(\Omega)}\}$$

and for r = 1 and r = 2

$$\tau_r^{-}(f, \Psi(n^{-1}))_{p(\Omega)} \le c\{E^{-}(f, H_{n-1})_{p(\Omega)} + \tau_r(f, \Psi(n^{-1}))_{p, p(\Omega)}\}.$$

In order to prove Theorem 1.3(II) we obtain some results for convex functions. Let $U \subset \mathbb{R}^d$ be a convex body. A function $f : U \to \mathbb{R}$ is called *almost midconvex* if $\sup\{\tilde{\Delta}^2_{\mathbf{h},U}f(\mathbf{x}) \mid \mathbf{h} \in \mathbb{R}^d\} = 0$ holds for every $\mathbf{x} \in U$ except a subset of U with a measure zero. As a corollary from the results in the Section 3 we get

Theorem 1.7 If a function $f : \Pi[\mathbf{a}; \mathbf{b}] \to \mathbb{R}$ is almost midconvex then f is equal almost everywhere to a convex function g and $f \ge g$.

2 A characterization of (1.3) in terms of best algebraic local approximation from below.

Here we use methods which are based on ideas of [2], [3] and [5] and prove Theorem 1.2. We start with some notations.

Let N be a fixed natural number. We set

$$\mathbb{Z} = \{0, 1, ..., N-1\}^d ; \ \mathbb{Z}' = \{0, 1, ..., N\}^d ; \ \mathbb{E} = \{0, 1\}^d;$$
$$z_k = \cos(\pi - \frac{k\pi}{N}), \ k = 0, 1, ..., N, \ z_{-1} = z_0 = -1, \ z_{N+1} = z_N = 1$$

For every $\mathbf{j} = (j_1, j_2, ..., j_d) \in \mathbb{Z}$ we denote

$$\boldsymbol{\Omega_{j}} = [z_{j_1}, \ z_{j_1+1}] \times \ \dots \ \times [z_{j_d}, \ z_{j_d+1}]$$

and for every $\mathbf{j} \in \mathbb{Z}'$ we denote

$$\mathbf{\Omega}'_{\mathbf{j}} = [z_{j_1-1}, \ z_{j_1+1}] \times \ \dots \ \times [z_{j_d-1}, \ z_{j_d+1}].$$

We set $\mu(v) = \int_0^v e^{\frac{-1}{u-u^2}} du / \int_0^1 e^{\frac{-1}{u-u^2}} du$ for 0 < v < 1, $\mu(v) = 0$ for $v \le 0$ and $\mu(v) = 1$ for $v \ge 1$. Therefore $\mu \in C^{\infty}(\mathbb{R})$ and we define

$$\mu_0(v) \stackrel{\text{def}}{=} 1 - \mu((v - z_0)/(z_1 - z_0)); \mu_s(v) \stackrel{\text{def}}{=} \mu((v - z_{s-1})/(z_s - z_{s-1}))(1 - \mu((v - z_s)/(z_{s+1} - z_s))) \quad for \ s = 1, 2, ..., N - 1; \mu_N(v) \stackrel{\text{def}}{=} \mu((v - z_{N-1})/(z_N - z_{N-1})).$$

Finally for every $\mathbf{j} \in \mathbb{Z}'$ we set $\mu_{\mathbf{j}}(\mathbf{x}) = \prod_{s=1}^d \mu_{j_s}(x_s)$. Therefore for every $\mathbf{x} \in \Omega$ we have

(2.1)
$$0 \le \mu_{\mathbf{j}}(\mathbf{x}) \le 1; \ \mu_{\mathbf{j}}(\mathbf{x}) = 0 \ if \ \mathbf{x} \notin \Omega_{\mathbf{j}}';$$

(2.2)
$$\sum_{\mathbf{j}\in\mathbb{Z}'}\mu_{\mathbf{j}}(\mathbf{x}) = 1.$$

In the statements below we collect some properties of the above quantities. Let $0 < t \le \frac{1}{2}$ and $N = \left\lceil \frac{2\pi}{t} \right\rceil + 1$. Then we have

- (2.3) $\Psi(t, \mathbf{x}) \le meas(U(t, \mathbf{x})) \le 2^d \Psi(t, \mathbf{x});$
- (2.4) $\Psi(t, \mathbf{x}) \sim \Psi(t, \mathbf{y}) \text{ for every } \mathbf{y} \in U(t, \mathbf{x});$
- (2.5) $\Psi(t, \mathbf{x}) \sim \Psi(t, \mathbf{x} + \mathbf{y}) \text{ for every } \mathbf{y} \in B(t, \mathbf{x});$
- (2.6) $c\Psi(t,\mathbf{x}) \le meas(\Omega'_{\mathbf{j}}) \le c\Psi(t,\mathbf{y}) \text{ for every } x, y \in \Omega'_{\mathbf{j}};$
- (2.7) $\Omega'_{\mathbf{j}} \subset U(t, \mathbf{x}) \quad for \quad any \quad \mathbf{x} \in \Omega'_{\mathbf{j}}.$

The inequalities (2.3), (2.4), (2.6) and (2.7) are proved in [3]. (2.5) follows from (2.3), (2.4) and definition of $B(t, \mathbf{x})$.

We prove first the following

Lemma 2.1 Let $0 < t \leq \frac{1}{2}$. Then for every $f \in L_p(\Omega)$ we have

(2.8)
$$\|\Psi(t,\cdot)^{-\frac{1}{p}}E^{-}(f,H_{r-1})_{p(U(t,\cdot))}\|_{p(\Omega)} \le cK_{r}^{-}(f,t)_{p};$$

(2.9)
$$K_r^-(f,t)_p \le c \|\Psi(t,\cdot)^{-\frac{1}{p}} E^-(f,H_{r-1})_{p(U(t,\cdot))}\|_{p(\Omega)}.$$

Proof. Let us begin with the proof of (2.9). We set

(2.10)
$$N = \left[\frac{2\pi}{t}\right] + 1$$

and use the notation for $\Omega_{\mathbf{j}}$, $\Omega'_{\mathbf{j}}$ and $\mu_{\mathbf{j}}$ from the beginning of Section 2. We denote by $Q_{\mathbf{j}} \in H_{r-1}$ the polynomial of best algebraic L_p approximation from below of degree r-1 to f in $\Omega'_{\mathbf{j}}$, $\mathbf{j} \in \mathbb{Z}'$. We set

(2.11)
$$g(\mathbf{x}) = \sum_{\mathbf{j} \in \mathbb{Z}'} \mu_{\mathbf{j}}(\mathbf{x}) Q_{\mathbf{j}}(\mathbf{x}).$$

From (2.10), (2.11), (2.3), (2.6) and (2.7) we obtain

$$(2.12) \|f - g\|_{p(\Omega)}^{p} = \|\sum_{\mathbf{j}\in\mathbb{Z}'}\mu_{\mathbf{j}}(f - Q_{\mathbf{j}})\|_{p(\Omega)}^{p}$$

$$\leq c\sum_{\mathbf{j}\in\mathbb{Z}'}\int_{\Omega_{\mathbf{j}}'}|f(\mathbf{x}) - Q_{\mathbf{j}}(\mathbf{x})|^{p}d\mathbf{x}$$

$$= c\sum_{\mathbf{j}\in\mathbb{Z}'}E^{-}(f, H_{r-1})_{p(\Omega_{\mathbf{j}}')}^{p}$$

$$= c\sum_{\mathbf{j}\in\mathbb{Z}'}meas(\Omega_{\mathbf{j}}')^{-1}\int_{\Omega_{\mathbf{j}}'}E^{-}(f, H_{r-1})_{p(U(t,\mathbf{x}))}^{p}d\mathbf{x}$$

$$\leq c\sum_{\mathbf{j}\in\mathbb{Z}'}meas(\Omega_{\mathbf{j}}')^{-1}\int_{\Omega_{\mathbf{j}}'}E^{-}(f, H_{r-1})_{p(U(t,\mathbf{x}))}^{p}d\mathbf{x}$$

$$\leq c\sum_{\mathbf{j}\in\mathbb{Z}'}f_{\Omega_{\mathbf{j}}'}(\Psi(t,\mathbf{x}))^{-1}E^{-}(f, H_{r-1})_{p(U(t,\mathbf{x}))}^{p}d\mathbf{x}$$

$$\leq c\int_{\Omega}(\Psi(t,\mathbf{x}))^{-1}E^{-}(f, H_{r-1})_{p(U(t,\mathbf{x}))}^{p}d\mathbf{x}$$

$$\leq c\|\Psi(t,\cdot)^{-\frac{1}{p}}E^{-}(f, H_{r-1})_{p(U(t,\cdot))}\|_{p(\Omega)}^{p}.$$

Fix α , $|\alpha| = r$ or $|\alpha| = l$. Let $\mathbf{x} \in \Omega_{\mathbf{j}}$, $\mathbf{j} \in \mathbb{Z}$. From the definitions of $\mu(\mathbf{x})$, $Q_{\mathbf{j}}(\mathbf{x})$ and $g(\mathbf{x})$ we have

$$g(\mathbf{x}) = Q_{\mathbf{j}}(\mathbf{x}) + \sum_{\epsilon \in \mathbb{E}} \mu_{\mathbf{j}+\epsilon}(\mathbf{x}) \left(Q_{\mathbf{j}+\epsilon}(\mathbf{x}) - Q_{\mathbf{j}}(\mathbf{x}) \right), \text{ where}$$
$$\mu_{\mathbf{j}+\epsilon}(\mathbf{x}) = \prod_{s; \epsilon_s=1} \mu \left(\frac{x_s - z_{j_s}}{z_{j_s+1} - z_{j_s}} \right) \cdot \prod_{s; \epsilon_s=0} \mu \left(1 - \left(\frac{x_s - z_{j_s}}{z_{j_s+1} - z_{j_s}} \right) \right)$$

•

and then from the last equality and $D^{\alpha}Q_{\mathbf{j}} = 0$, it follows that

$$D^{\alpha}g(\mathbf{x}) = \sum_{\epsilon \in \mathbb{E}} \sum_{0 \le \beta \le \alpha} {\alpha \choose \beta} D^{\alpha-\beta} \mu_{\mathbf{j}+\epsilon}(\mathbf{x}) D^{\beta} \left(Q_{\mathbf{j}+\epsilon}(\mathbf{x}) - Q_{\mathbf{j}}(\mathbf{x}) \right)$$

Now using (2.6), (2.7), the definitions of $\mu_{\mathbf{j}}$, $Q_{\mathbf{j}}$ and \mathbb{E} and Markov's inequality ($(b-a)^i \|g^{(i)}\|_{p[a,b]} \leq c(r) \|g\|_{p[a,b]}$ for $g \in H_r$) we have

$$\begin{split} \|\Psi^{\alpha}(t)D^{\alpha}g\|_{p(\Omega_{\mathbf{j}})} &\leq c\Psi^{\alpha}(t,z_{\mathbf{j}})\|D^{\alpha}g\|_{p(\Omega_{\mathbf{j}})} \\ &\leq c\Psi^{\alpha}(t,z_{\mathbf{j}})\sum_{\epsilon\in\mathbb{E}}\sum_{\mathbf{0}\leq\beta\leq\alpha} \binom{\alpha}{\beta}\|D^{\alpha-\beta}\mu_{\mathbf{j}+\epsilon}\|_{\infty(\Omega_{\mathbf{j}})}\|D^{\beta}\left(Q_{\mathbf{j}+\epsilon}-Q_{\mathbf{j}}\right)\|_{p(\Omega_{\mathbf{j}})} \\ &\leq c\Psi^{\alpha}(t,z_{\mathbf{j}})\sum_{\epsilon\in\mathbb{E}}\sum_{\mathbf{0}\leq\beta\leq\alpha}\prod_{s=1}^{d}\frac{\|\mu^{(|\alpha-\beta|)}\|_{\infty[0,1]}}{|z_{j_{s}+1}-z_{j_{s}}|^{\alpha_{s}-\beta_{s}}}\|D^{\beta}\left(Q_{\mathbf{j}+\epsilon}-Q_{\mathbf{j}}\right)\|_{p(\Omega_{\mathbf{j}})} \\ &\leq c\sum_{\epsilon\in\mathbb{E}}\sum_{\mathbf{0}\leq\beta\leq\alpha}\prod_{s=1}^{d}|z_{j_{s}+1}-z_{j_{s}}|^{\beta_{s}}\|D^{\beta}\left(Q_{\mathbf{j}+\epsilon}-Q_{\mathbf{j}}\right)\|_{p(\Omega_{\mathbf{j}})} \\ &\leq c\sum_{\epsilon\in\mathbb{E}}\|Q_{\mathbf{j}+\epsilon}-Q_{\mathbf{j}}\|_{p(\Omega_{\mathbf{j}})} \end{split}$$

$$\leq c \sum_{\epsilon \in \mathbb{E}} \left(\|f - Q_{\mathbf{j}+\epsilon}\|_{p(\Omega_{\mathbf{j}})} + \|f - Q_{\mathbf{j}}\|_{p(\Omega_{\mathbf{j}})} \right)$$

$$\leq c E^{-}(f, H_{r-1})_{p(\Omega_{\mathbf{j}}')}^{p}.$$

Hence

$$(2.13) \|\Psi^{\alpha}(t)D^{\alpha}g\|_{p(\Omega)}^{p} \leq c \sum_{\mathbf{j}\in\mathbb{Z}'} E^{-}(f,H_{r-1})_{p(\Omega'_{\mathbf{j}})}^{p} \\ = c \sum_{\mathbf{j}\in\mathbb{Z}'} meas(\Omega'_{\mathbf{j}})^{-1} \int_{\Omega'_{\mathbf{j}}} E^{-}(f,H_{r-1})_{p(\Omega'_{\mathbf{j}})}^{p} d\mathbf{x} \\ \leq c \sum_{\mathbf{j}\in\mathbb{Z}'} meas(\Omega'_{\mathbf{j}})^{-1} \int_{\Omega'_{\mathbf{j}}} E^{-}(f,H_{r-1})_{p(U(t,\mathbf{x}))}^{p} d\mathbf{x} \\ \leq c \sum_{\mathbf{j}\in\mathbb{Z}'} \int_{\Omega'_{\mathbf{j}}} (\Psi(t,\mathbf{x}))^{-1} E^{-}(f,H_{r-1})_{p(U(t,\mathbf{x}))}^{p} d\mathbf{x} \\ \leq c \int_{\Omega} (\Psi(t,\mathbf{x}))^{-\frac{1}{p}} E^{-}(f,H_{r-1})_{p(U(t,\mathbf{x}))}^{p} d\mathbf{x} \\ \leq c \|\Psi(t,\cdot)^{-\frac{1}{p}} E^{-}(f,H_{r-1})_{p(U(t,\mathbf{x}))}^{p} d\mathbf{x} \\ \leq c \|\Psi(t,\cdot)^{-\frac{1}{p}} E^{-}(f,H_{r-1})_{p(U(t,\cdot))}\|_{p(\Omega)}^{p}.$$

In this way (2.9) follows from (1.3), (2.12) and (2.13).

We turn our attention to (2.8).

Let $\alpha = (\alpha_1, ..., \alpha_d), |\alpha| = r$ be multi-index and $\mathbf{z} = (z_1, ..., z_d) \in \mathbb{R}^d$. We define

 $|\mathbf{z}^{\alpha}| = \prod_{i=1}^{d} |z_i|^{\alpha_i}$

Let $\Pi = \Pi[\mathbf{a}; \mathbf{b}]$ and let $g \in W_p^l(\Pi)$. As a corollary from Theorem 2 and Theorem 1 in [3] we get

(2.14)
$$E^{-}(g, H_{r-1})_{p(\Pi)} \leq c \sum_{|\alpha|=r,l} |(\mathbf{b} - \mathbf{a})^{\alpha}| \| D^{\alpha}g \|_{p(\Pi)}.$$

Let g be any function in $W^l_p(\Omega), g(\mathbf{x}) \leq f(\mathbf{x})$, $\mathbf{x} \in \Omega$. Then we have (note that $Q \leq g$ implies $Q \leq f$)

$$E^{-}(f, H_{r-1})_{p(U(t,\mathbf{x}))} \le E^{-}(g, H_{r-1})_{p(U(t,\mathbf{x}))} + ||f - g||_{p(U(t,\mathbf{x}))}$$

and hence

(2.15)
$$\|\Psi(t,\cdot)^{-\frac{1}{p}}E^{-}(f,H_{r-1})_{p(U(t,\cdot))}\|_{p(\Omega)}$$

$$\leq \|\Psi(t,\cdot)^{-\frac{1}{p}}E^{-}(g,H_{r-1})_{p(U(t,\cdot))}\|_{p(\Omega)} + \|\Psi(t,\cdot)^{-\frac{1}{p}}\|f-g\|_{p(U(t,\cdot))}\|_{p(\Omega)}.$$

Using (2.14), (2.3) and (2.4) we obtain

(2.16)
$$E^{-}(g, H_{r-1})_{p(U(t,\mathbf{x})))} \leq c \sum_{|\alpha|=r,l} \Psi^{\alpha}(t, \mathbf{x}) \|D^{\alpha}g\|_{p(U(t,\mathbf{x}))}$$
$$\leq c \sum_{|\alpha|=r,l} \|\Psi^{\alpha}(t, \cdot)D^{\alpha}g\|_{p(U(t,\mathbf{x}))}.$$

From Lemma 4 in [3] we have that

(2.17)
$$\|\Psi^{-\frac{1}{p}}(t,\cdot)\|G\|_{p(U(t,\cdot))}\|_{p(\Omega)} \le c\|G\|_{p(\Omega)}$$

for $G \in L_p(\Omega)$ and $t \in (0, \frac{1}{2}]$. Then from (2.16) and (2.17) we get

$$\|\Psi^{-\frac{1}{p}}(t,\cdot)\|f - g\|_{p(U(t,\cdot))}\|_{p(\Omega)} \le c\|f - g\|_{p(\Omega)};$$
$$\|\Psi^{-\frac{1}{p}}(t,\cdot)\|E^{-}(g,H_{r-1})\|_{p(U(t,\cdot))}\|_{p(\Omega)} \le c\sum_{|\alpha|=r,l} \|\Psi^{\alpha}(t,\cdot)D^{\alpha}g\|_{p(\Omega)}.$$

Hence using (2.15) we get

$$\|\Psi^{-\frac{1}{p}}(t,\cdot)\|E^{-}(f,H_{r-1})\|_{p(U(t,\cdot))}\|_{p(\Omega)} \le c \left\{\|f-g\|_{p(\Omega)} + \sum_{|\alpha|=r,l} \|\Psi^{\alpha}(t)D^{\alpha}g\|_{p(\Omega)}\right\}.$$

Taking an infimum on all $g \in W_p^r(\Omega)$, $g \leq f$ in the above inequality we prove (2.8).

Proof of Theorem 1.2. We have to investigate only the case $t \in (\frac{1}{2}, 1]$, because for $t \in (0, \frac{1}{2}]$ Theorem 1.2 is equal to Lemma 2.1. Let $t \in (\frac{1}{2}, 1]$. Then from the definitions of $\Psi(t, \mathbf{x})$ and $U(t, \mathbf{x})$ it follows that

$$\begin{split} \|\Psi^{-\frac{1}{p}}(t,\cdot)E^{-}(f,H_{r-1})_{p(U(t,\cdot))}\|_{p(\Omega)} &\leq 4^{d}E^{-}(f,H_{r-1})_{p(\Omega)} \\ &\leq c\|\Psi^{-\frac{1}{p}}(\frac{1}{2},\cdot)E^{-}(f,H_{r-1})_{p(U(t,\cdot))}\|_{p(\Omega)} \leq c\|\Psi^{-\frac{1}{p}}(t,\cdot)E^{-}(f,H_{r-1})_{p(U(t,\cdot))}\|_{p(\Omega)} \end{split}$$

Hence from Lemma 2.1(2.8) with $t = \frac{1}{2}$ and the monotonicity of the K-functional (1.3) with respect to t we get

$$\begin{split} \|\Psi^{-\frac{1}{p}}(t,\cdot)E^{-}(f,H_{r-1})_{p(U(t,\cdot))}\|_{p(\Omega)} &\leq cK_{r}^{-}(f,\frac{1}{2})_{p} \leq cK_{r}^{-}(f,t)_{p};\\ K_{r}^{-}(f,t)_{p} \leq cE^{-}(f,H_{r-1})_{p(\Omega)} &\leq c\|\Psi^{-\frac{1}{p}}(t,\cdot)E^{-}(f,H_{r-1})_{p(U(t,\cdot))}\|_{p(\Omega)}. \end{split}$$

3 Whitney-type theorems for best approximations from below.

We make use of some properties of the moduli which follows immediately from the definition

(3.1)
$$\tau_r^-(f,U)_p \leq r \|f\|_{p(U)} \text{ for } r = 1, \ 2 \quad \text{if } f(\mathbf{x}) \geq 0 \text{ for every } \mathbf{x} \in U;$$

(3.2)
$$\tau_2^-(f,U)_p = 0 \quad if \quad f \quad is \quad convex \quad on \quad U \quad ;$$

(3.3)
$$\tau_r^-(f+g,U)_p \le \tau_r^-(f,U)_p + \tau_r^-(g,U)_p \quad for \ r=1, \ 2;$$

(3.4)
$$\tau_2^-(f_+, U)_p \le \tau_2^-(f, U)_p, \text{ where } f(\mathbf{x})_+ \stackrel{\text{def}}{=} \begin{cases} f(\mathbf{x}) & \text{if } f(\mathbf{x}) \ge 0; \\ 0 & \text{otherwise.} \end{cases}$$

Remark 3.1. In (3.1) and (3.3) we use $\sup\{\tilde{\Delta}_{\mathbf{h},U}^r f(\mathbf{x}) \mid \mathbf{h} \in \mathbb{R}^d\} \geq \tilde{\Delta}_{\mathbf{0},U}^r f(\mathbf{x}) = 0$. **Remark 3.2.** $\tau_r^-(f-g,U)_p \leq \tau_r^-(f,U)_p + \tau_r^-(g,U)_p$ is not true in general. For example d = 1, r = 2, U = [-1,1], f(x) = const and $g(x) = x^2$. **Remark 3.3.** In (3.4) we use that if $f(\mathbf{x}) \geq 0$ then

$$\sup\{\tilde{\Delta}_{\mathbf{h},U}^2 f(\mathbf{x})_+ \mid \mathbf{h} \in \mathbb{R}^d\} = \sup\{2f(\mathbf{x}) - f(\mathbf{x} + \mathbf{h})_+ - f(\mathbf{x} - \mathbf{h})_+ \mid \mathbf{h} \in \mathbb{R}^d\}$$
$$\leq \sup\{\tilde{\Delta}_{\mathbf{h},U}^2 f(\mathbf{x}) \mid \mathbf{h} \in \mathbb{R}^d\}$$

and if $f(\mathbf{x}) < 0$ then $\sup\{\tilde{\Delta}_{\mathbf{h},U}^2 f(\mathbf{x})_+ \mid \mathbf{h} \in \mathbb{R}^d\} = 0 \le \sup\{\tilde{\Delta}_{\mathbf{h},U}^2 f(\mathbf{x}) \mid \mathbf{h} \in \mathbb{R}^d\}.$

Proof of Theorem 1.3(I). The statement (I) of Theorem 1.3 is similar to Theorem 4.1 from [2] and the proof is the same. Let $M = \inf\{f(\mathbf{y}) \mid \mathbf{y} \in \Pi = \Pi[\mathbf{a}; \mathbf{b}]\}$. Then $E^{-}(f, H_0)_{p(\Pi)} = \|f - M\|_{p(\Pi)}$ and for every $\mathbf{x} \in \Pi$ we have

$$f(\mathbf{x}) - M = f(\mathbf{x}) - \inf\{f(\mathbf{y}) \mid \mathbf{y} \in \Pi\}$$

= sup{ $f(\mathbf{x}) - f(\mathbf{x} + \mathbf{h}) \mid \mathbf{x} + \mathbf{h} \in \Pi\}$
= sup{ $\tilde{\Delta}^{1}_{\mathbf{h},\Pi} f(\mathbf{x}) \mid \mathbf{h} \in \mathbb{R}^{d}$ }.

Taking L_p norm in this inequality we prove the lemma.

Now we turn our attention to the case r = 2 (Theorem 1.3 (II)). We start with some lemmas which are conected with the best multivariate algebraic approximations from below of convex functions.

Lemma 3.1 Let $E \subset \Pi[\mathbf{a}; \mathbf{b}] \subset \mathbb{R}^d$ be an open convex body with a measure $\mu(E) < \frac{1}{2^d d!} \mu(\Pi[\mathbf{a}; \mathbf{b}])$. Then there are $k \in \{1, ..., d\}$ and measurable subsets E_{k-} and E_{k+} and for every $\mathbf{x} \in E$ there exist $s(\mathbf{x}) \in \partial E$ and $\mathbf{y}(\mathbf{x}) \in \Pi[\mathbf{a}; \mathbf{b}] \setminus (E \cup \partial E)$, such that

1)
$$E = E_{k-} \cup E_{k+};$$

2) If $\mathbf{x} \in E_{k*}$ (* = + or -), $\mathbf{y}(\mathbf{x}) - \mathbf{s}(\mathbf{x}) = \mathbf{s}(\mathbf{x}) - \mathbf{x} = t_{\mathbf{x}}e_k$, where e_k is the "k-th" unit coordinate vector and $sign(t_{\mathbf{x}}) = *$.

Proof. Let $\mathbf{x} \in E$ and $k \in \{1, ..., d\}$. We define

$$E_k(\mathbf{x}) \stackrel{\text{def}}{=} \{ \mathbf{z} \in E \mid z_i = x_i \; \forall i = 1, ..., d \;, \; i \neq k \} ,$$
$$m_{k,E}^{-}(\mathbf{x}) \stackrel{\text{def}}{=} \inf \{ z_k \mid \mathbf{z} \in E_k(\mathbf{x}) \}$$

and

$$m_{k,E}^+(\mathbf{x}) \stackrel{\text{def}}{=} \sup \{ z_k \mid \mathbf{z} \in E_k(\mathbf{x}) \}$$

From $\mu(E) < \frac{1}{2^d d!} \mu(\Pi[\mathbf{a}; \mathbf{b}])$ and convexity of E we have that there exists $k \in \{1, ..., d\}$ such that $m_{k,E}^+(\mathbf{x}) - m_{k,E}^-(\mathbf{x}) < \frac{|b_k - a_k|}{2}$ for every $\mathbf{x} \in E$. (If we assume that for every $k \in \{1, ..., d\}$

there exist $\mathbf{x}(k) \in E$ such that $m_{k,E}^+(\mathbf{x}(k)) - m_{k,E}^-(\mathbf{x}(k)) \ge \frac{|b_k - a_k|}{2}$ then from convexity of E we have that $\mu(E) \ge \frac{1}{2^d d!} \mu(\Pi[\mathbf{a}; \mathbf{b}])$.) Thus we reduce the problem for $E \subset \mathbb{R}^d$ to the problem for $E_k(\mathbf{x}) \subset \mathbb{R}^1$. We define

$$c_{k,E}(\mathbf{x}) \stackrel{\text{def}}{=} \begin{cases} m_{k,E}^{+}(\mathbf{x}) & \text{if } \frac{b_{k}+a_{k}}{2} \leq m_{k,E}^{-}(\mathbf{x}) \\ m_{k,E}^{-}(\mathbf{x}) + m_{k,E}^{+}(\mathbf{x}) - \frac{b_{k}+a_{k}}{2} & \text{if } \frac{b_{k}+a_{k}}{2} \in \left(m_{k,E}^{-}(\mathbf{x}), m_{k,E}^{+}(\mathbf{x})\right) \\ m_{k,E}^{-}(\mathbf{x}) & \text{if } \frac{b_{k}+a_{k}}{2} \geq m_{k,E}^{+}(\mathbf{x}). \end{cases}$$

We set $E_{k-} = \left\{ \mathbf{z} \in E \mid z_k \in (m_{k,E}^-(\mathbf{z}), c_{k,E}(\mathbf{z})], E_{k+} = \left\{ \mathbf{z} \in E \mid z_k \in (c_{k,E}(\mathbf{z}), m_{k,E}^+(\mathbf{z})) \right\}$ and let $s(\mathbf{x}) = (s(\mathbf{x})_1, ..., s(\mathbf{x})_d)$ be such that

$$s(\mathbf{x})_i \stackrel{\text{def}}{=} \begin{cases} x_i & \text{if } i \neq k \\ m_{k,E}^*(\mathbf{x}) & \text{if } i = k \text{ and } \mathbf{x} \in E_{k*}, \ * = + \text{ or } - \end{cases}$$

The functions $m_{k,E}^-(\mathbf{x})$ and $m_{k,E}^+(\mathbf{x})$ are continuous because they are face functions of the convex body E. Then from the construction the subsets E_{k-} and E_{k+} have continuous boundary and then they are measurable. Also from the construction of the subsets E_{k-} and E_{k+} we have $E = E_{k-} \cup E_{k+}, \mathbf{y}(\mathbf{x}) = 2\mathbf{s}(\mathbf{x}) - \mathbf{x} \in \Pi[\mathbf{a}; \mathbf{b}] \setminus (E \cup \partial E)$ and if $\mathbf{x} \in E_{k*}$ (* = + or -) then $\mathbf{y}(\mathbf{x}) - \mathbf{s}(\mathbf{x}) = \mathbf{s}(\mathbf{x}) - \mathbf{x} = t_{\mathbf{x}}e_k$, where e_k is the "k-th" unit coordinate vector and $sign(t_{\mathbf{x}}) = *$.

Lemma 3.2 Let f be convex on $\Pi[\mathbf{0}; \mathbf{a}], f(\mathbf{x}) \ge 0, f(\mathbf{0}) = 0$ and $p \in [1, \infty)$. Then

$$||f||_{p(\Pi[\mathbf{0};\mathbf{a}])} \le cE(f,H_0)_{p(\Pi[\mathbf{0};\mathbf{a}])}$$

Proof. Let M be such that $E^{-}(f, H_{0})_{p} = ||f - M||_{p}$. If M = 0 (which is possible for example when p = 1 and f vanishes in a set E with a measure $\mu(E) \geq \frac{1}{2}\mu(\Pi[\mathbf{0}; \mathbf{a}])$), then the statement of Lemma 3.2 holds as an equality with constant c = 1. In the other cases (when M > 0) we set $E_{*} = \{\mathbf{x} \in \Pi[\mathbf{0}; \mathbf{a}] \mid sign(f(\mathbf{x}) - M) = *\}$, where * = "-" or "+".

Let $E_0 = \partial E_-$. For $\mathbf{x} \in E_-$ we define $g(\mathbf{x}) \stackrel{\text{def}}{=} t_{\mathbf{x}} M$, where $\mathbf{x} = t_{\mathbf{x}} \mathbf{y}$ and $\mathbf{y} \in E_0$. Since $\int_{E_-} (M - f)^p \ge \int_{E_-} (M - g)^p$ and

$$\mu \left\{ \mathbf{x} \in E_{-} \mid (1 - t_{\mathbf{x}})^{p} \ge \theta \right\} = \mu \left\{ \mathbf{x} \in E_{-} \mid t_{\mathbf{x}} \le (1 - \theta^{\frac{1}{p}}) \right\} = \mu(E_{-})(1 - \theta^{\frac{1}{p}})^{d},$$

we have

(3.5)
$$\int_{E_{-}} (M - f(\mathbf{x}))^{p} d\mathbf{x} \geq \int_{E_{-}} (M - g(\mathbf{x}))^{p} d\mathbf{x}$$
$$= M^{p} \int_{E_{-}} (1 - t_{\mathbf{x}})^{p} d\mathbf{x}$$
$$= M^{p} \int_{E_{-}} \int_{0}^{(1 - t_{\mathbf{x}})^{p}} 1 d\theta d\mathbf{x}$$
$$= M^{p} \int_{0}^{1} \int_{(1 - \theta^{\frac{1}{p}})^{d} E_{-}} 1 d\mathbf{x} d\theta$$

$$= M^{p} \mu(E_{-}) \int_{0}^{1} (1 - \theta^{\frac{1}{p}})^{d} d\theta$$

= $M^{p} \mu(E_{-}) p \int_{0}^{1} t^{d} (1 - t)^{p-1} dt$
= $\binom{p+d}{d}^{-1} \int_{E_{-}} M^{p}.$

From (3.5) and the trivial inequality $\int_{E_{-}} f^p \leq \int_{E_{-}} M^p$ we get

(3.6)
$$\int_{E_{-}} f^{p} \leq \binom{p+d}{d} \int_{E_{-}} (M-f)^{p}$$

f - M and M are positive numbers in E_+ . Then the convexity of x^p gives

(3.7)
$$\int_{E_+} f^p \le 2^{p-1} \int_{E_+} (f - M)^p + 2^{p-1} \int_{E_+} M^p.$$

If we assume that $\mu(E_{-}) < \frac{1}{2^{d}d!}\mu(\Pi[\mathbf{0}; \mathbf{a}])$ then from Lemma 3.1 (with $E = E_{-}$) and the convexity of f we have that $M - f(\mathbf{x}) \leq f(\mathbf{y}(\mathbf{x})) - M$ ($f(\mathbf{s}(\mathbf{x})) = M$) for any $\mathbf{x} \in E_{-}$. Take the power p-1 in the both sides of the last inequality and integrating on $\mathbf{x} \in E_{-}$ we obtain

$$\int_{E_{-}} (M - f(\mathbf{x}))^{p-1} d\mathbf{x} \le \int_{E_{-}} (f(\mathbf{y}(\mathbf{x})) - M)^{p-1} d\mathbf{x} < \int_{E_{+}} (f(\mathbf{y}) - M)^{p-1} d\mathbf{y}.$$

But this is a contradiction, because from the characterization of the element of best approximation (see [7]) we have

$$\int_{\Pi[\mathbf{0};\mathbf{a}]} (f(\mathbf{x}) - M)^{p-1} sign(f(\mathbf{x}) - M) d\mathbf{x} = 0.$$

Therefore $\mu(E_{-}) \geq \frac{1}{2^{d}d!}\mu(\Pi[\mathbf{0};\mathbf{a}])$ i.e. $\mu(E_{+}) \leq (2^{d}d! - 1)\mu(E_{-})$. Then using the last one, (3.5), (3.6) and (3.7) we obtain

(3.8)
$$\int_{E_{+}} f^{p} \leq 2^{p-1} \int_{E_{+}} (f-M)^{p} + 2^{p-1} \binom{p+d}{d} (2^{d}d! - 1) \int_{E_{-}} (M-f)^{p}.$$

Now from (3.6) and (3.8) it follows

$$||f||_{p(\Pi[\mathbf{0};\mathbf{a}])} \le c||f - M||_{p(\Pi[\mathbf{0};\mathbf{a}])}.$$

Lemma 3.3 If f is convex in $\Pi[\mathbf{a}; \mathbf{b}]$ then $E^{-}(f, H_1)_{p(\Pi[\mathbf{a}; \mathbf{b}])} \leq c E(f, H_1)_{p(\Pi[\mathbf{a}; \mathbf{b}])}$.

Proof. Without loss of generality we may assume that $f(\mathbf{y}) = \liminf_{\mathbf{x}\to\mathbf{y}} f(\mathbf{x})$ for every boundary point \mathbf{y} of $\Pi[\mathbf{a}; \mathbf{b}]$. Let $Q(\mathbf{x})$ be the first degree polynomial of best L_p approximation

to f. The function f - Q is convex and let its minimum be achieved (from the above assumtion) at the point $\mathbf{u} \in \Pi[\mathbf{a}; \mathbf{b}]$. Applying Lemma 3.2 to the convex function

$$g(\mathbf{x}) = f(\mathbf{x}) - Q(\mathbf{x}) - (f(\mathbf{u}) - Q(\mathbf{u})),$$

separately on the parallelepipeds $\Pi[\mathbf{u}; \alpha \mathbf{a} + (\mathbf{1} - \alpha)\mathbf{b}]$, where $\alpha = (\alpha_1 \dots \alpha_d) \in \{0, 1\}^d$ we have

$$\|g\|_{p(\Pi[\mathbf{u};\alpha\mathbf{a}+(1-\alpha)\mathbf{b}])} \le c \ E(g, H_0)_{p(\Pi[\mathbf{u};\alpha\mathbf{a}+(1-\alpha)\mathbf{b}])} \le c \|g+(f(\mathbf{u})-Q(\mathbf{u}))\|_{p(\Pi[\mathbf{u};\alpha\mathbf{a}+(1-\alpha)\mathbf{b}])}$$

for every $\alpha \in \{0,1\}^d$.

Now adding the above inequalities raised to the power p and then take the power $\frac{1}{p}$ of the sum we obtain

$$\|f - Q - (f(\mathbf{u}) - Q(\mathbf{u}))\|_{p(\Pi[\mathbf{a};\mathbf{b}])} \le c \|f - Q\|_{p(\Pi[\mathbf{a};\mathbf{b}])} = c \ E(f, H_1)_{p(\Pi[\mathbf{a};\mathbf{b}])}.$$

But $f(\mathbf{x}) \ge Q(\mathbf{x}) + f(\mathbf{u}) - Q(\mathbf{u})$ and $Q(x) + f(\mathbf{u}) - Q(\mathbf{u}) \in H_1$. Then

$$E^{-}(f, H_{1})_{p(\Pi[\mathbf{a};\mathbf{b}])} \leq ||f - Q - (f(\mathbf{u}) - Q(\mathbf{u}))||_{p(\Pi[\mathbf{a};\mathbf{b}])} \leq c E(f, H_{1})_{p(\Pi[\mathbf{a};\mathbf{b}])}.$$

In order to get the result for the approximations from below by linear functions we need some statements which are more complificated than in case d = 1.

Let $A = \mathbf{A}_1 \dots \mathbf{A}_{d+1}$ be d-dimensional simplex. If $\mathbf{x} \in A$, then $\mathbf{x} = \sum_{i=1}^{d+1} \alpha_i(\mathbf{x}) \mathbf{A}_i$, where $\alpha_i(\mathbf{x}) \ge 0$ and $\sum_{i=1}^{d+1} \alpha_i(\mathbf{x}) = 1$. $\alpha_1(\mathbf{x}), \dots, \alpha_{d+1}(\mathbf{x})$ are usually called *barycentric coordinates* of \mathbf{x} with respect to $\mathbf{A}_1, \dots, \mathbf{A}_{d+1}$.

For n > d and $i = 1 \dots n - 1$ we consider the following subsets of A

(3.9)
$$M_{k,i}^n = \{ \mathbf{x} \in A \mid \alpha_k(\mathbf{x}) \ge \frac{i}{n} \}.$$

Let $i_k(\mathbf{x}) = [\alpha_k(\mathbf{x})n]$, i.e. $\alpha_k(\mathbf{x}) \in [\frac{i_k(\mathbf{x})}{n}, \frac{i_k(\mathbf{x})+1}{n}]$. Using that $\sum_{k=1}^{d+1} \alpha_k(\mathbf{x}) = 1$ we obtain

(3.10)
$$\sum_{k=1}^{d+1} i_k(\mathbf{x}) \in [n-d, n].$$

Lemma 3.4 Let $A = \mathbf{A}_1 \dots \mathbf{A}_{d+1}$ be d-dimensional simplex on \mathbb{R}^d , $p \in [1, \infty)$, $f \in L_p(A)$ be non-negative in A and $f(\mathbf{A}_i) = 0$ for every $i = 1, \dots, d+1$. Then

$$||f||_{p(A)} \le c\tau_2^-(f,A)_p$$

Proof. Let n > d and i = 1, ..., n - 1 be integer. We define

(3.11)
$$D_{\mathbf{x}}^{n,i}f(\mathbf{y}) \stackrel{\text{def}}{=} -if(\mathbf{y}) + nf(\mathbf{x}) - (n-i)f(\mathbf{y} + \frac{n}{n-i}(\mathbf{x} - \mathbf{y})).$$

We need the following inequality

(3.12)
$$D_{\mathbf{x}}^{n,i}f^{p}(\mathbf{A}_{k}) \leq (D_{\mathbf{x}}^{n,i}f(\mathbf{A}_{k}))_{+}^{p} \ for \ \mathbf{x} \in M_{k,i}^{n}.$$

From (3.11) and $f(\mathbf{A}_k) = 0$ it follows that

$$f(\mathbf{x}) = \frac{1}{n} D_{\mathbf{x}}^{n,i} f(\mathbf{A}_k) + \frac{i}{n} f(\mathbf{A}_k) + \frac{n-i}{n} f(\mathbf{A}_k + \frac{n}{n-i} (\mathbf{x} - \mathbf{A}_k))$$
$$= \frac{1}{n} D_{\mathbf{x}}^{n,i} f(\mathbf{A}_k) + \frac{i-1}{n} f(\mathbf{A}_k) + \frac{n-i}{n} f(\mathbf{A}_k + \frac{n}{n-i} (\mathbf{x} - \mathbf{A}_k))$$

Hence the trivial inequality $D_{\mathbf{x}}^{n,i}f(\mathbf{A}_k) \leq (D_{\mathbf{x}}^{n,i}f(\mathbf{A}_k))_+$ and convexity of x^p for $p \geq 1$ give

$$f^{p}(\mathbf{x}) \leq \left(\frac{1}{n}(D_{\mathbf{x}}^{n,i}f(\mathbf{A}_{k}))_{+} + \frac{i-1}{n}f(\mathbf{A}_{k}) + \frac{n-i}{n}f(\mathbf{A}_{k} + \frac{n}{n-i}(\mathbf{x}-\mathbf{A}_{k}))\right)^{p}$$
$$\leq \frac{1}{n}(D_{\mathbf{x}}^{n,i}f(\mathbf{A}_{k}))_{+}^{p} + \frac{i-1}{n}f^{p}(\mathbf{A}_{k}) + \frac{n-i}{n}f^{p}(\mathbf{A}_{k} + \frac{n}{n-i}(\mathbf{x}-\mathbf{A}_{k})).$$

Also from (3.11) and $f^p(\mathbf{A}_k) = 0$ it follows that

$$f^{p}(\mathbf{x}) = \frac{1}{n} D_{\mathbf{x}}^{n,i} f^{p}(\mathbf{A}_{k}) + \frac{i}{n} f^{p}(\mathbf{A}_{k}) + \frac{n-i}{n} f^{p}(\mathbf{A}_{k} + \frac{n}{n-i}(\mathbf{x} - \mathbf{A}_{k}))$$
$$= \frac{1}{n} D_{\mathbf{x}}^{n,i} f^{p}(\mathbf{A}_{k}) + \frac{i-1}{n} f^{p}(\mathbf{A}_{k}) + \frac{n-i}{n} f^{p}(\mathbf{A}_{k} + \frac{n}{n-i}(\mathbf{x} - \mathbf{A}_{k})).$$

The last two inequalities prove (3.12). Using (3.9), (3.10) and (3.12) we derive

$$(3.13) \qquad \sum_{k=1}^{d+1} \sum_{i=1}^{n-1} \int_{M_{k,i}^{n}} \left(D_{\mathbf{x}}^{n,i} f(\mathbf{A}_{k}) \right)_{+}^{p} d\mathbf{x} \\ \geq \sum_{k=1}^{d+1} \sum_{i=1}^{n-1} \int_{M_{k,i}^{n}} nf^{p}(\mathbf{x}) - (n-i)f^{p}(\mathbf{A}_{k} + \frac{n}{n-i}(\mathbf{x} - \mathbf{A}_{k}))d\mathbf{x} \\ = n \left(\sum_{k=1}^{d+1} \sum_{i=1}^{n-1} \int_{M_{k,i}^{n}} f^{p}(\mathbf{x})d\mathbf{x} - (d+1) \sum_{i=1}^{n-1} \left(\frac{n-i}{n} \right)^{d+1} \int_{A} f^{p}(\mathbf{x})d\mathbf{x} \right) \\ = n \left(\sum_{k=1}^{d+1} \sum_{i=1}^{n-1} i \int_{M_{k,i}^{n} \setminus M_{k,i+1}^{n}} f^{p}(\mathbf{x})d\mathbf{x} - (d+1) \sum_{i=1}^{n-1} \left(\frac{i}{n} \right)^{d+1} \int_{A} f^{p}(\mathbf{x})d\mathbf{x} \right) \\ = n \left(\sum_{k=1}^{d+1} \sum_{i=1}^{n-1} i \int_{M_{k,i}^{n} \setminus M_{k,i+1}^{n}} f^{p}(\mathbf{x})d\mathbf{x} - (d+1) n \sum_{i=1}^{n-1} \left(\frac{i}{n} \right)^{d+1} \int_{A} f^{p}(\mathbf{x})d\mathbf{x} \right) \\ \geq n \left((n-d) - (d+1)n \int_{0}^{1} t^{d+1} dt \right) \int_{A} f^{p}(\mathbf{x})d\mathbf{x} \\ = n \frac{n-d(d+2)}{d+2} \int_{A} f^{p}(\mathbf{x})d\mathbf{x}.$$

Here in third line we use that the map $\mathbf{x} \mapsto \mathbf{A}_k + \frac{n}{n-i}(\mathbf{x} - \mathbf{A}_k)$ is affine with a center on \mathbf{A}_k and stretchs the simplex $M_{k,i}^n$ to the simplex A.

Immediately from (3.11) (the definition of $D_{\mathbf{x}}^{n,i}f(\mathbf{y})$) we have

$$D_{\mathbf{x}}^{n,1}f(\mathbf{y}) = \sum_{j=1}^{n-1} j\tilde{\Delta}_{\frac{\mathbf{x}-\mathbf{y}}{n-1}}^2 f(\mathbf{y}+j\frac{\mathbf{x}-\mathbf{y}}{n-1})$$

and

$$D_{\mathbf{x}}^{n,i}f(\mathbf{y}) = iD_{\mathbf{x}}^{n-i+1,1}f(\mathbf{y}) + (n-i)D_{\mathbf{y}+\frac{n-i+1}{n-i}(\mathbf{x}-\mathbf{y})}^{i,1}f(\mathbf{y}+\frac{n}{n-i}(\mathbf{x}-\mathbf{y}))$$

Hence

$$D_{\mathbf{x}}^{n,i}f(\mathbf{y}) = i\sum_{s=1}^{n-i}s\tilde{\Delta}_{\frac{\mathbf{x}-\mathbf{y}}{n-i}}^2f(\mathbf{y}+s\frac{\mathbf{x}-\mathbf{y}}{n-i}) + (n-i)\sum_{s=n-i+1}^{n-1}(n-s)\tilde{\Delta}_{\frac{\mathbf{x}-\mathbf{y}}{n-i}}^2f(\mathbf{y}+s\frac{\mathbf{x}-\mathbf{y}}{n-i}).$$

Using the above equality, (3.12) and the definition (1.6) we obtain

$$\begin{aligned} (3.14) \quad & \left\{ \sum_{k=1}^{d+1} \sum_{i=1}^{n-1} \int_{M_{k,i}^{n}} \left(D_{\mathbf{x}}^{n,i} f(\mathbf{A}_{k}) \right)_{+}^{p} d\mathbf{x} \right\}^{\frac{1}{p}} \\ & \leq \sum_{k=1}^{d+1} \sum_{i=1}^{n-1} \left\{ \int_{M_{k,i}^{n}} \left(D_{\mathbf{x}}^{n,i} f(\mathbf{A}_{k}) \right)_{+}^{p} d\mathbf{x} \right\}^{\frac{1}{p}} \\ & \leq \sum_{k=1}^{d+1} \sum_{i=1}^{n-1} \left(i \sum_{s=1}^{n-i} s \left\{ \int_{M_{k,i}^{n}} \left(\tilde{\Delta}_{\frac{\mathbf{x}-\mathbf{A}_{k}}{n-i}}^{\mathbf{x}-\mathbf{A}_{k}} f(\mathbf{A}_{k} + s \frac{\mathbf{x}-\mathbf{A}_{k}}{n-i}) \right)_{+}^{p} d\mathbf{x} \right\}^{\frac{1}{p}} \\ & + (n-i) \sum_{s=n-i+1}^{n-1} (n-s) \left\{ \int_{M_{k,i}^{n}} \left(\tilde{\Delta}_{\frac{\mathbf{x}-\mathbf{A}_{k}}{n-i}}^{\mathbf{x}-\mathbf{A}_{k}} f(\mathbf{A}_{k} + s \frac{\mathbf{x}-\mathbf{A}_{k}}{n-i}) \right)_{+}^{p} d\mathbf{x} \right\}^{\frac{1}{p}} \\ & \leq \sum_{k=1}^{d+1} \sum_{i=1}^{n-1} \left(i \sum_{s=1}^{n-i} s \left\{ \int_{M_{k,i}^{n}} \left(\sup\{ \tilde{\Delta}_{\mathbf{h},A}^{2} f(\mathbf{A}_{k} + s \frac{\mathbf{x}-\mathbf{A}_{k}}{n-i}) \right) + \mathbf{h} \in \mathbb{R}^{d} \right\} \right)^{p} d\mathbf{x} \right\}^{\frac{1}{p}} \\ & + (n-i) \sum_{s=n-i+1}^{n-1} (n-s) \left\{ \int_{M_{k,i}^{n}} \left(\sup\{ \tilde{\Delta}_{\mathbf{h},A}^{2} f(\mathbf{A}_{k} + s \frac{\mathbf{x}-\mathbf{A}_{k}}{n-i}) \right) + \mathbf{h} \in \mathbb{R}^{d} \right\} \right)^{p} d\mathbf{x} \right\}^{\frac{1}{p}} \\ & = \sum_{k=1}^{d+1} \sum_{i=1}^{n-1} \left(i \sum_{s=1}^{n-i} s \left\{ \int_{M_{k,i}^{n}} \left(\sup\{ \tilde{\Delta}_{\mathbf{h},A}^{2} f(\mathbf{A}_{k} + s \frac{\mathbf{x}-\mathbf{A}_{k}}{n-i}) \right) + \mathbf{h} \in \mathbb{R}^{d} \right\} \right)^{p} d\mathbf{x} \right\}^{\frac{1}{p}} \\ & + (n-i) \sum_{s=n-i+1}^{n-1} (n-s) \left(\frac{n-i}{s} \right)^{\frac{d}{p}} \left\{ \int_{M_{k,n}^{n}} \left(\sup\{ \tilde{\Delta}_{\mathbf{h},A}^{2} f(\mathbf{x}) + \mathbf{h} \in \mathbb{R}^{d} \right\} \right)^{p} d\mathbf{x} \right\}^{\frac{1}{p}} \\ & \leq \left[\sum_{k=1}^{d+1} \sum_{i=1}^{n-1} \left(i \sum_{s=1}^{n-i} s \left(\frac{n-i}{s} \right)^{\frac{d}{p}} + (n-i) \sum_{s=n-i+1}^{n-i} (n-s) \left(\frac{n-i}{s} \right)^{\frac{d}{p}} \right) \right] \\ & \leq \left[\sum_{k=1}^{d+1} \sum_{i=1}^{n-i} \left(i \sum_{s=1}^{n-i} s \left(\frac{n-i}{s} \right)^{\frac{d}{p}} + (n-i) \sum_{s=n-i+1}^{n-i} (n-s) \left(\frac{n-i}{s} \right)^{\frac{d}{p}} \right) \right] \\ & \times \left\{ \int_{A} \left(\sup\{ \tilde{\Delta}_{\mathbf{h},A}^{2} f(\mathbf{x}) + \mathbf{h} \in \mathbb{R}^{d} \} \right)^{p} d\mathbf{x} \right\}^{\frac{1}{p}} \\ & \leq c_{n} \tau_{2}^{-} (f, A)_{p}. \end{aligned}$$

The inequalites (3.13) and (3.14) with $n = (d+1)^2$ prove the lemma.

Let $U \subset \mathbb{R}^d$ be a polytope and let $f \in L_p(U)$ $(p \in [1, \infty))$ be bounded from below. We set

(3.15)
$$C_U f(\mathbf{x}) \stackrel{\text{def}}{=} \inf \left\{ \sum_{i=1}^{d+1} \alpha_i f(\mathbf{x}_i) \mid \mathbf{x} = \sum_{i=1}^{d+1} \alpha_i \mathbf{x}_i, \sum_{i=1}^{d+1} \alpha_i = 1, \ \alpha_i \ge 0, \ \mathbf{x}_i \in U, \ i = 1, ..., d+1 \right\}.$$

Immediately from (3.15) and [6] we have

(3.16) $C_U f$ is convex on U, continuous on every open subset of U and Lipschitz function on every compact subset of the interior of U;

(3.17) If h is convex and is majorized by f on U then $h(\mathbf{x}) \leq C_U f(\mathbf{x})$ for all $\mathbf{x} \in U$, i.e. $C_U f$ is the biggest convex minorant of f in U.

For $g: U \to \mathbb{R}$, let *epigraph* of g be the set $epi(g) = \{(\mathbf{x}, t) \in \mathbb{R}^{d+1} \mid \mathbf{x} \in U, t \ge g(\mathbf{x})\}$. We say $\mathbf{x} \in U$ is *extreme with respect to the convex function* g if $g(\mathbf{x}) < \infty$ and g is not linear on any relatively open segment containing \mathbf{x} . Hence \mathbf{x} is extreme with respect to g if and only if $(\mathbf{x}, g(\mathbf{x}))$ is an extreme point of epi(g).

Let $EP(g) \subset U$ be the set of extreme points with respect to the convex function g. Then from (3.15) and (3.17) it follows that

(3.18) for any positive ϵ and for any $\mathbf{x} \in EP(C_U f)$ there exists $\mathbf{y} \in U$, such that $\|\mathbf{x} - \mathbf{y}\| < \epsilon$ and $|f(\mathbf{y}) - C_U f(\mathbf{x})| < \epsilon$.

Lemma 3.5 Let $\Pi = \Pi[\mathbf{a}; \mathbf{b}]$ and $f \in L_p(\Pi)$ ($p \in [1, \infty)$) be bounded from below. Then

$$||f - C_{\Pi}f||_{p(\Pi)} \le \tau_2^-(f,\Pi)_p.$$

Proof. We denote with $\mathbf{A}_1, ..., \mathbf{A}_{2^d}$ the vertices of Π . From (3.15) it follows that $\inf\{f(\mathbf{y}) \mid \mathbf{y} \in \Pi\} \leq C_{\Pi} f(\mathbf{x}) \leq f(\mathbf{x})$ for every $\mathbf{x} \in \Pi$ and hence $f - C_{\Pi} f \in L_p(\Pi)$.

Let ϵ be arbitrary positive. From the absolute continuity of the Lebesgue integral there is a positive $\eta < \frac{1}{2} \min \{\max\{a_i; b_i\} - \min\{a_i; b_i\} \mid i = 1, ..., d\}$, such that

(3.19)
$$||f - C_{\Pi}f||_{p(\Pi \setminus \Pi(\eta))} \le \epsilon,$$

where $\Pi(\eta) = \Pi[\mathbf{c}; \mathbf{h}], c_i \stackrel{\text{def}}{=} \min\{a_i; b_i\} + \eta \text{ and } h_i \stackrel{\text{def}}{=} \max\{a_i; b_i\} - \eta \text{ for every } i = 1, ..., d.$

The set $\Pi(\eta)$ is compact subset of the interior of Π and then (3.16) gives that there exists a positive constant L such that

(3.20)
$$|C_{\Pi}f(\mathbf{x}) - C_{\Pi}f(\mathbf{y})| \le L \|\mathbf{x} - \mathbf{y}\| \text{ for any two points } \mathbf{x}, \ \mathbf{y} \in \Pi(\eta).$$

For i = 1, ..., d we set $n_i \stackrel{\text{def}}{=} \left[\epsilon \frac{|h_i - c_i|}{3L} \right] + 1$ and let $\mathbb{Z}(\epsilon) \stackrel{\text{def}}{=} \left\{ \mathbf{j} \in \mathbb{Z}^d \mid j_i \in [0, n_i], i = 1, ..., d \right\}$. For every $\mathbf{j} \in \mathbb{Z}(\epsilon)$ we set $\mathbf{z_j} \stackrel{\text{def}}{=} \left(c_1 + j_1 \frac{(h_1 - c_1)}{n_1}, ..., c_d + j_d \frac{(h_d - c_d)}{n_d} \right) \in \Pi(\eta)$. As a corollary from the theorem of J.-C. Aggeri (Krein-Milman's type theorem for convex functions (see [1])) we derive that for every $\mathbf{x} \in \Pi(\eta)$ and every $\delta > 0$ there exist points $\mathbf{y}_i(\mathbf{x}, \delta) \in EP(C_{\Pi}f)$ i = 1, ..., d + 1 with $\mathbf{x} = \sum_{i=1}^{d+1} \alpha_i \mathbf{y}_i(\mathbf{x}, \delta)$, $\sum_{i=1}^{d+1} \alpha_i = 1, \ \alpha_i \ge 0$ for which $C_{\Pi}f(\mathbf{x}) \ge \sum_{i=1}^{d+1} \alpha_i C_{\Pi}f(\mathbf{y}_i(\mathbf{x}, \delta)) - \delta$. Let $EP(\epsilon) \stackrel{\text{def}}{=} \{\mathbf{y}_i(\mathbf{z}_j, \frac{\epsilon}{3}) \mid i = 1, ..., d + 1, \ \mathbf{j} \in \mathbb{Z}(\epsilon)\}$.

This is a m-points set where $m \leq (d+1) \prod_{i=1}^{d} (n_i+1)$.

We define

$$s_1(\mathbf{x}) \stackrel{\text{def}}{=} \min\left\{\sum_{i=1}^{d+1} \alpha_i C_{\Pi} f(\mathbf{a}_i) \mid \mathbf{x} = \sum_{i=1}^{d+1} \alpha_i \mathbf{a}_i, \sum_{i=1}^{d+1} \alpha_i = 1, \ \alpha_i \ge 0, \ \mathbf{a}_i \in EP(\epsilon)\right\}.$$

This is a first degree convex interpolation spline for $C_{\Pi}f$ with knots in $EP(\epsilon)$.

For every $\mathbf{x} \in \Pi(\eta)$ we have that there exist set of points $\left\{\mathbf{z}_{\mathbf{j}(\mathbf{x},i)}\right\}_{i=1}^{d+1}$, such that $\mathbf{x} = \sum_{i=1}^{d+1} \alpha_i \mathbf{z}_{\mathbf{j}(\mathbf{x},i)}, \sum_{i=1}^{d+1} \alpha_i = 1, \ \alpha_i \ge 0, \ \mathbf{j}(\mathbf{x},i) \in \mathbb{Z}(\epsilon) \text{ and } |\mathbf{z}_{\mathbf{j}(\mathbf{x},i)} - \mathbf{x}| \le \frac{\epsilon}{3L} \text{ for } i = 1, ..., d+1$ For this points (3.20) gives

$$|C_{\Pi}f(\mathbf{x}) - C_{\Pi}f(\mathbf{z}_{\mathbf{j}(\mathbf{x},i)})| \le \frac{\epsilon}{3} \quad \forall i = 1, ..., d+1$$

Using the definitions of $s_1(\mathbf{x})$, $EP(\epsilon)$, the points $\{\mathbf{z}_{\mathbf{j}(\mathbf{x},i)}\}_{i=1}^{d+1}$ and the last inequality we obtain

$$0 \leq s_{1}(\mathbf{x}) - C_{\Pi}f(\mathbf{x}) \leq \sum_{i=1}^{d+1} \alpha_{i} \sum_{k=1}^{d+1} \alpha_{i,k} C_{\Pi}f\left(\mathbf{y}_{k}\left(\mathbf{z}_{\mathbf{j}(\mathbf{x},i)}, \frac{\epsilon}{3}\right)\right) - C_{\Pi}f(\mathbf{x})$$
$$\leq \sum_{i=1}^{d+1} \alpha_{i}\left(C_{\Pi}f(\mathbf{z}_{\mathbf{j}(\mathbf{x},i)}) + \frac{\epsilon}{3}\right) - C_{\Pi}f(\mathbf{x})$$
$$\leq \sum_{i=1}^{d+1} \alpha_{i}\left((C_{\Pi}f(\mathbf{x}) + \frac{\epsilon}{3}) + \frac{\epsilon}{3}\right) - C_{\Pi}f(\mathbf{x})$$
$$\leq \epsilon \frac{2}{3}$$

Using s_1 and (3.18) we can find a first degree interpolation spline $s(\mathbf{x})$ with knots $\{\mathbf{y}_1, \ldots, \mathbf{y}_m\} \in \Pi$ such that $C_{\Pi}f(\mathbf{x}) \leq s(\mathbf{x}) \leq C_{\Pi}f(\mathbf{x}) + \epsilon$ for $\mathbf{x} \in \Pi(\eta)$ and $f(\mathbf{y}_i) = s(\mathbf{y}_i)$, $i = 1, \ldots, m$.

Suppose $\Pi(\eta) \subset \bigcup_{i=1}^{k} D_i$ where $D_i = \mathbf{y}_{i_1} \dots \mathbf{y}_{i_{d+1}}$ are d-dimensional simplecies with $\{\mathbf{y}_{i_1}, \dots, \mathbf{y}_{i_{d+1}}\} \subset \{\mathbf{y}_1, \dots, \mathbf{y}_m\}$ and the restrictions of $s(\mathbf{x})$ on D_i are affine functions. Then from (3.19)

(3.21)
$$\|f - C_{\Pi}f\|_{p(\Pi)}^{p} = \|f - C_{\Pi}f\|_{p(\Pi(\eta))}^{p} + \|f - C_{\Pi}f\|_{p(\Pi\setminus\Pi(\eta))}^{p}$$
$$\leq \sum_{i=1}^{k} \|f - C_{\Pi}f\|_{p(D_{i}\cap\Pi(\eta))}^{p} + \epsilon^{p}.$$

Using the definitions of $C_{\Pi}f$ and s, trivial equality $(f - s) = (f - s)_+ - (s - f)_+$, applying Lemma 3.4 to $(f - s)_+$ in D_i and properties (3.1), (3.2) and (3.4), we get

$$(3.22) ||f - C_{\Pi}f||_{p(D_i \cap \Pi(\eta))} \leq ||f - s||_{p(D_i \cap \Pi(\eta))} + ||s - C_{\Pi}f||_{p(D_i \cap \Pi(\eta))}$$

$$\leq \|(f-s)_{+}\|_{p(D_{i}\cap\Pi(\eta))} + \|(s-f)_{+}\|_{p(D_{i}\cap\Pi(\eta))} + \epsilon\mu(D_{i}\cap\Pi(\eta))^{\frac{1}{p}} \\ \leq \|(f-s)_{+}\|_{p(D_{i})} + \|(s-f)_{+}\|_{p(D_{i}\cap\Pi(\eta))} + \epsilon\mu(D_{i}\cap\Pi(\eta))^{\frac{1}{p}} \\ \leq c\tau_{2}^{-}((f-s)_{+}, D_{i})_{p} + \|s-C_{\Pi}f\|_{p(D_{i}\cap\Pi(\eta))} + \epsilon\mu(D_{i}\cap\Pi(\eta))^{\frac{1}{p}} \\ \leq c\tau_{2}^{-}((f-s), D_{i})_{p} + 2\epsilon\mu(D_{i}\cap\Pi(\eta))^{\frac{1}{p}} \\ \leq c\tau_{2}^{-}((f-s), D_{i})_{p} + 2\epsilon\mu(D_{i})^{\frac{1}{p}}.$$

The inequalities (3.21) and (3.22) give

$$\|f - C_{\Pi}f\|_{p(\Pi)} \leq c \left\{ \sum_{i=1}^{k} (\tau_{2}^{-}(f, D_{i})_{p} + 2\epsilon\mu(D_{i})^{\frac{1}{p}})^{p} \right\}^{\frac{1}{p}} + \epsilon$$
$$\leq c \left\{ \sum_{i=1}^{k} \tau_{2}^{-}(f, D_{i})_{p}^{p} \right\}^{\frac{1}{p}} + 2\epsilon (\sum_{i=1}^{k} \mu(D_{i}))^{\frac{1}{p}} + \epsilon$$
$$\leq c\tau_{2}^{-}(f, \Pi)_{p} + (2\mu(\Pi) + 1)\epsilon$$
$$\leq c(\tau_{2}^{-}(f, \Pi)_{p} + \epsilon).$$

Lemma 3.5 is proved.

Lemma 3.6 Under the assumtion of Lemma 3.5 we have

$$E^{-}(f, H_1)_{p(\Pi)} \sim E(f, H_1)_{p(\Pi)} + \tau_2^{-}(f, \Pi)_p.$$

Proof. The inequality $C_{\Pi}f \leq f$ implies

$$E^{-}(f, H_1)_{p(\Pi)} \le E^{-}(C_{\Pi}f, H_1)_{p(\Pi)} + \|f - C_{\Pi}f\|_{p(\Pi)}.$$

Lemma 3.3 applied to $C_{\Pi}f$ gives

$$E^{-}(C_{\Pi}f, H_{1})_{p(\Pi)} \leq cE(C_{\Pi}f, H_{1})_{p(\Pi)}.$$

Combining the above two inequalities and

$$E(C_{\Pi}f, H_1)_{p(\Pi)} \le E(f, H_1)_{p(\Pi)} + \|f - C_{\Pi}f\|_{p(\Pi)}$$

we prove the direct inequality in view of Lemma 3.5. In order to get the other direction of the equivalence in Lemma 3.6 we estimate both terms of its right-hand side by $E^{-}(f, H_1)_{p(\Pi)}$. Obviously $E(f, H_1)_{p(\Pi)} \leq E^{-}(f, H_1)_{p(\Pi)}$. Let $Q \in H_1$ be such that $f \geq Q$ and $E^{-}(f, H_1)_{p(\Pi)} = ||f - Q||_{p(\Omega)}$. From the property (3.1) we have

$$\tau_2^-(f,\Pi)_p = \tau_2^-(f-Q,\Pi)_p \le 2||f-Q||_{p(\Pi)} = 2E^-(f,H_1)_{p(\Pi)}.$$

Hence

$$E(f, H_1)_{p(\Pi)} + \tau_2^-(f, \Pi)_p \le 3E^-(f, H_1)_{p(\Pi)}.$$

Proof of Theorem 1.3(II). Using that $E(f, H_{r-1})_{p(\Pi)} \sim \omega_r(f, \Pi)_p$ (see[4]), as a corollary from the last lemma we obtain Theorem 1.3(II). Here the result for $p = \infty$ is trivial.

Proof of Theorem 1.7. From the definition we have that the almost midconvex function is bounded from above. Then Theorem 1.7 follows from Lemma 3.5. \Box

4 Main results.

Proof of Theorem 1.5.

Here we use the ideas from [1]. Utilizing Theorem 1.3(I) and Theorem 1.2 we obtain a characterization of $K^{-}(f, \Psi(t), L_p, W_p^1, W_p^{l_1})$. We demonstrate the proof in the more complicated case- r = 2.

Using $K_2^-(f,\rho t) \leq max\{1,\rho^{l_2}\}K_2^-(f,t)$ for $\rho > 0$, Theorem 1.2, Theorem 1.3(II), Theorem 1.4 (with $r = 2, \pi = \frac{1}{2}B(t, \mathbf{x}), \Pi = U(t, \mathbf{x})$ and R = 2), (1.8) and (1.9) we have

$$(4.1) K_{2}^{-}(f,t)_{p} \sim K_{2}^{-}(f,\rho t)_{p} \\ \sim \|\Psi(t,\cdot)^{-\frac{1}{p}}E^{-}(f,H_{r-1})_{p(U(t,\cdot))}\|_{p(\Omega)} \\ \sim \|\Psi(t,\cdot)^{-\frac{1}{p}}\{\omega_{2}(f,U(\rho t,\cdot))_{p}+\tau_{2}^{-}(f,U(\rho t,\cdot))_{p}\}\|_{p(\Omega)} \\ \sim \|\Psi(t,\cdot)^{-\frac{1}{p}}\{\tau_{2}(f,\frac{1}{2}B(\rho t,\cdot))_{p,p(U(\rho t,\cdot))}+\tau_{2}^{-}(f,U(\rho t,\cdot))_{p}\}\|_{p(\Omega)} \\ \sim \left\|\Psi(t,\cdot)^{-\frac{1}{p}}\left\{\int_{U(\rho t,\cdot)}\Psi(\rho t,\cdot)^{-1}\int_{\frac{1}{2}B(\rho t,\cdot)}|\Delta_{\mathbf{v},U(\rho t,\cdot)}^{2}f(\mathbf{y})|^{p}d\mathbf{v}d\mathbf{y}\right\}^{\frac{1}{p}}\right\|_{p(\Omega)} \\ + \left\|\Psi(t,\cdot)^{-\frac{1}{p}}\left\{\int_{U(\rho t,\cdot)}\left[\sup\{\tilde{\Delta}_{\mathbf{h},U(\rho t,\cdot)}^{2}f(\mathbf{y})\mid \mathbf{h}\in\mathbb{R}^{d}\}\right]^{p}d\mathbf{y}\right\}^{\frac{1}{p}}\right\|_{p(\Omega)}.$$

From (2.4) with d = 1 and the definitions of $\Psi(t, \mathbf{x})$ and $U(t, \mathbf{x})$ (see also [2], Lemma 3) we have

(4.2)
$$\Psi(\frac{1}{6}t, \mathbf{x}) \le \Psi(t, \mathbf{y}) \quad for \quad every \quad \mathbf{y} \in U(\frac{1}{6}t, \mathbf{x}) \quad and \quad \mathbf{x} \in \Omega \quad and$$

(4.3)
$$\Psi(t, \mathbf{y}) \le \Psi(4t, \mathbf{x}) \quad for \quad every \quad \mathbf{y} \in U(t, \mathbf{x}) \quad and \quad \mathbf{x} \in \Omega.$$

Then using (2.3), (2.4), (4.2) and Lemma 2.1 we get

$$\begin{split} & \left\| \Psi(t,\cdot)^{-\frac{1}{p}} \left\{ \int_{U(\frac{1}{6}t,\cdot)} \Psi(\frac{1}{6}t,\cdot)^{-1} \int_{\frac{1}{2}B(\frac{1}{6}t,\cdot)} |\Delta^{2}_{\mathbf{v},U(\frac{1}{6}t,\cdot)}f(\mathbf{y})|^{p} d\mathbf{v} d\mathbf{y} \right\}^{\frac{1}{p}} \right\|_{p(\Omega)} \\ & \leq c \left\| \Psi(t,\cdot)^{-\frac{1}{p}} \left\{ \int_{U(\frac{1}{6}t,\cdot)} \Psi(t,\mathbf{y})^{-1} \int_{\frac{1}{2}B(t,\mathbf{y})} |\Delta^{2}_{\mathbf{v},U(t,\mathbf{y})}f(\mathbf{y})|^{p} d\mathbf{v} d\mathbf{y} \right\}^{\frac{1}{p}} \right\|_{p(\Omega)} \\ & \sim \left\| \Psi(\frac{1}{6}t,\cdot)^{-\frac{1}{p}} \left\{ \int_{U(\frac{1}{6}t,\cdot)} \Psi(t,\mathbf{y})^{-1} \int_{\frac{1}{2}B(t,\mathbf{y})} |\Delta^{2}_{\mathbf{v},U(t,\mathbf{y})}f(\mathbf{y})|^{p} d\mathbf{v} d\mathbf{y} \right\}^{\frac{1}{p}} \right\|_{p(\Omega)} \\ & \sim \tau_{2}(f,\Psi(t))_{p,p(\Omega)}. \end{split}$$

Using the same arguments we have that

$$\left\|\Psi(t,\cdot)^{-\frac{1}{p}}\left\{\int_{U(\frac{1}{6}t,\cdot)}\sup\{\tilde{\Delta}_{\mathbf{h},U(\frac{1}{6}t,\cdot)}^{2}f(\mathbf{y}) \mid \mathbf{h}\in\mathbb{R}^{d}\}^{p}d\mathbf{y}\right\}^{\frac{1}{p}}\right\|_{p(\Omega)} \leq c\tau_{2}^{-}(f,\Psi(t))_{p(\Omega)}$$

Then from (4.1) with $\rho = \frac{1}{6}$ we get the inequality

$$K^{-}(f, \Psi(t), L_{p}, W_{p}^{2}, W_{p}^{l_{2}}) \leq c\{\tau_{2}^{-}(f, \Psi(t))_{p(\Omega)} + \tau_{2}(f, \Psi(t))_{p,p(\Omega)}\}.$$

The proof of the opposite inequality is the same. Using Lemma 2.1, (4.3) and (2.4) we get

$$\begin{split} &\tau_{2}(f,\Psi(t))_{p,p(\Omega)}\\ &\sim \left\|\Psi(\frac{1}{2}t,\cdot)^{-\frac{1}{p}}\left\{\int_{U(\frac{1}{2}t,\cdot)}\Psi(t,\mathbf{y})^{-1}\int_{\frac{1}{2}B(t,\mathbf{y})}|\Delta^{2}_{\mathbf{v},U(t,\mathbf{y})}f(\mathbf{y})|^{p}d\mathbf{v}d\mathbf{y}\right\}^{\frac{1}{p}}\right\|_{p(\Omega)}\\ &\leq c\left\|\Psi(t,\cdot)^{-\frac{1}{p}}\left\{\int_{U(\frac{1}{6}t,\cdot)}\Psi(t,\mathbf{y})^{-1}\int_{\frac{1}{2}B(t,\mathbf{y})}|\Delta^{2}_{\mathbf{v},U(t,\mathbf{y})}f(\mathbf{y})|^{p}d\mathbf{v}d\mathbf{y}\right\}^{\frac{1}{p}}\right\|_{p(\Omega)}\\ &\leq c\left\|\Psi(t,\cdot)^{-\frac{1}{p}}\left\{\int_{U(4t,\cdot)}\Psi(4t,\cdot)^{-1}\int_{\frac{1}{2}B(4t,\cdot)}|\Delta^{2}_{\mathbf{v},U(4t,\cdot)}f(\mathbf{y})|^{p}d\mathbf{v}d\mathbf{y}\right\}^{\frac{1}{p}}\right\|_{p(\Omega)} \end{split}$$

In the same way we have that

$$\tau_2^-(f,\Psi(t))_{p(\Omega)} \le c \left\| \Psi(t,\cdot)^{-\frac{1}{p}} \left\{ \int_{U(4t,\cdot)} \sup\{\tilde{\Delta}^2_{\mathbf{h},U(4t,\cdot)} f(\mathbf{y}) \mid \mathbf{h} \in \mathbb{R}^d \}^p d\mathbf{y} \right\}^{\frac{1}{p}} \right\|_{p(\Omega)}$$

Then from (4.1) with $\rho = 4$ we get the inequality

$$\tau_2^{-}(f, \Psi(t))_{p(\Omega)} + \tau_2(f, \Psi(t))_{p, p(\Omega)} \le cK^{-}(f, \Psi(t), L_p, W_p^2, W_p^{l_2}).$$

The following result for the unconstrained K-functional is valid (see [4] Theorem 1.3).

Lemma 4.1

$$c\tau_r(f,\Psi(t))_{p,p(\Omega)} \le K\left(f,\Psi(t),L_p,W_p^r\right) \le c\tau_r(f,\Psi(t))_{p,p(\Omega)}$$

Proof of Theorem 1.6. Applying Theorem 1.5 together with Theorem 1.1 and Lemma 4.1 we obtain Theorem 1.6. \Box

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