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# Greedy Wavelet Projections are Bounded on BV \*

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## Abstract

Let  $BV = BV(\mathbb{R}^d)$  be the space of functions of bounded variation on  $\mathbb{R}^d$  with  $d \geq 2$ . Let  $\psi_\lambda$ ,  $\lambda \in \Delta$ , be a wavelet basis of compactly supported functions normalized in BV, i.e.  $|\psi_\lambda|_{BV(\mathbb{R}^d)} = 1$ ,  $\lambda \in \Delta$ . Each  $f \in BV$  has a unique wavelet expansion  $\sum_{\lambda \in \Delta} c_\lambda(f) \psi_\lambda$  with convergence in  $L_1(\mathbb{R}^d)$ . If  $\Lambda_N(f)$  is the set of  $N$  indices  $\lambda \in \Delta$  for which  $|c_\lambda(f)|$  are largest (with ties handled in an arbitrary way), then  $\mathcal{G}_N(f) := \sum_{\lambda \in \Lambda_N(f)} c_\lambda(f) \psi_\lambda$  is called a greedy approximation to  $f$ . It is shown that  $|\mathcal{G}_N(f)|_{BV(\mathbb{R}^d)} \leq C|f|_{BV(\mathbb{R}^d)}$  with  $C$  a constant independent of  $f$ . This answers in the affirmative a conjecture of Meyer [15] (see p. 79).

**AMS subject classification:** 42C40, 46B70, 26B35, 42B25.

**Key Words:**  $N$ -term approximation, greedy approximation, functions of bounded variation, thresholding, bounded projections.

## 1 Introduction

The space  $BV := BV(\Omega)$  of functions of bounded variation on a domain  $\Omega \subset \mathbb{R}^d$  is important in mathematics (geometric measure theory, differential geometry) and applications (image processing, nonlinear PDEs). The structure of BV is complicated by the fact that neither it nor the closely related Sobolev space  $W^1(L_1(\Omega))$  have an unconditional basis (BV does not even have a basis). Wavelet decompositions of BV functions, while not characterizing this space, give fine information (see [4, 19, 2]) about its structure and these decompositions can be used to solve various extremal problems.

Consider, for example, the extremal problem

$$K(f, t) := K(f, t; L_2(\Omega), BV(\Omega)) := \inf_{g \in BV(\Omega)} \|f - g\|_{L_2(\Omega)} + t|g|_{BV(\Omega)}, \quad (1.1)$$

where  $\Omega = [0, 1]^2$  and  $t > 0$  is a parameter. The expression (1.1) is called a K-functional in interpolation of linear operators. It is used to describe interpolation spaces between  $L_2(\Omega)$

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and  $BV(\Omega)$ . This and related functionals also occur in image processing in such problems as denoising and deblurring. The rate of decay of  $K(f, t)$  as  $t \rightarrow 0$  gives information about the smoothness of  $f$  relative to  $L_2(\Omega)$  and  $BV(\Omega)$ . A function  $g = g_t$  is called a near minimizer (with constant  $C$ ) to (1.1) if

$$\|f - g_t\|_{L_2(\Omega)} + t|g_t|_{BV(\Omega)} \leq CK(f, t).$$

One would like simple constructive methods for finding minimizers or near minimizers to (1.1).

In [4], it is shown that thresholding the Haar decomposition of  $f$  provides a near minimizer to (1.1). Namely, if  $H_\lambda$ ,  $\lambda \in \Delta$ , is the Haar basis on  $[0, 1]^2$ , then given  $f \in L_2([0, 1]^2)$ , we can write

$$f = \sum_{\lambda \in \Delta} c_\lambda(f) H_\lambda$$

with the  $H_\lambda$  normalized in  $L_2([0, 1]^2)$  (which is equivalent to normalizing in  $BV([0, 1]^2)$ ). For each  $t > 0$ , a near minimizer  $g_t$  is given by thresholding the Haar series:

$$g_t := T_{t^2} f := \sum_{\lambda \in \Lambda(f, t^2)} c_\lambda(f) H_\lambda,$$

where for any  $t > 0$

$$\Lambda(f, t) := \{\lambda : |c_\lambda(f)| > t\}.$$

The proof that thresholding is a near minimizer relies on three basic results concerning Haar decompositions and  $BV$ . To describe these, we introduce the concept of  $N$ -term approximation using the Haar basis. We define  $\Sigma_N^w$  as the collection of all functions  $S = \sum_{\lambda \in \Lambda} c_\lambda H_\lambda$ , where  $\Lambda \subset \Delta$  is any index set with cardinality  $\#(\Lambda) \leq N$ . Given  $f \in L_2([0, 1]^2)$ , we consider the approximation of  $f$  using the elements of  $\Sigma_N^w$ :

$$\sigma_N(f)_{L_2([0, 1]^2)} := \inf_{S \in \Sigma_N^w} \|f - S\|_{L_2([0, 1]^2)}.$$

The first of these basic results is the following direct estimate (see [4]) for the approximation error:

$$\sigma_N(f)_{L_2([0, 1]^2)} \leq C_0 N^{-1/2} |f|_{BV([0, 1]^2)}, \quad N = 1, 2, \dots$$

This inequality is called an inequality of Jackson type (corresponding to analogous inequalities in approximation by algebraic polynomials). The Jackson inequality is proved by showing that the Haar coefficients of a  $BV([0, 1]^2)$  function are in weak  $\ell_1$ . That is

$$\#(\Lambda(f, \epsilon)) \leq C_0 \epsilon^{-1}, \quad \epsilon > 0.$$

This weak  $\ell_1$  property was shown in [6] to hold in the more general setting of wavelet expansions of functions in  $BV(\mathbb{R}^d)$  using compactly supported orthogonal wavelets. This allows the generalization of the Jackson inequality to arbitrary space dimensions and arbitrary compactly supported orthogonal wavelet systems (see Lemma 4.2).

The second basic result (see [4]) is the Bernstein inequality which (in the case of  $[0, 1]^2$ ) says that

$$|S|_{\text{BV}([0,1]^2)} \leq C_0 N^{1/2} \|S\|_{L_2([0,1]^2)}, \quad S \in \Sigma_N^w, \quad N = 1, 2, \dots$$

We shall show in §5 that this inequality also generalizes to  $\mathbb{R}^d$  and general compactly supported orthogonal wavelet systems.

The Jackson and Bernstein inequalities are not enough to show that thresholding the Haar expansion is an approximate minimizer for (1.1). One also needs the stability of thresholding in  $\text{BV}([0, 1]^2)$ :

$$|T_\epsilon(f)|_{\text{BV}([0,1]^2)} \leq C_0 |f|_{\text{BV}([0,1]^2)}, \quad f \in \text{BV}([0, 1]^2).$$

This remarkable property says that projecting onto any sum involving the  $N$  largest wavelet coefficients of the Haar series of a function in  $\text{BV}([0, 1]^2)$  results in a function with controllable  $\text{BV}([0, 1]^2)$  norm. Note that this property does not hold for projecting onto an arbitrary  $N$ -term sum of the Haar series nor does it hold in  $\mathbb{R}^1$  (see §7). This stability result for Haar expansions was generalized to space dimensions  $d > 2$  in [19]. Yves Meyer [15] (see p. 79) has conjectured that this property holds for any compactly supported wavelet system. The main result of this paper is to prove this conjecture.

**Theorem 1.1** *Let  $\varphi$  be a compactly supported univariate scaling function in  $\text{BV}(\mathbb{R}^1)$  which generates the compactly supported orthogonal wavelet  $\psi$ . For  $d \geq 2$ , we consider the multivariate orthogonal wavelet system  $(\psi_\lambda)_{\lambda \in \Delta}$  obtained from  $\varphi$  and  $\psi$ , and normalized in  $\text{BV}(\mathbb{R}^d)$ . Then this wavelet system has the following BV stability property. If  $f \in \text{BV}(\mathbb{R}^d)$ ,  $d \geq 2$ , let*

$$f = \sum_{\lambda \in \Delta} c_\lambda(f) \psi_\lambda$$

*be the wavelet expansion of  $f$ . Let for any  $N$ ,  $\Lambda_N(f)$  be the set of  $N$  indices  $\lambda \in \Delta$  for which  $|c_\lambda(f)|$  are largest. Then the nonlinear operator*

$$\mathcal{G}_N(f) := \sum_{\lambda \in \Lambda_N(f)} c_\lambda(f) \psi_\lambda$$

*satisfies*

$$|\mathcal{G}_N(f)|_{\text{BV}(\mathbb{R}^d)} \leq C(\varphi, d) |f|_{\text{BV}(\mathbb{R}^d)}.$$

As a consequence of this theorem we shall also show that  $\mathcal{G}_N(f)$  realizes the K-functional for the pair  $(L_{d^*}(\mathbb{R}^d), \text{BV}(\mathbb{R}^d))$ .

**Theorem 1.2** *Let  $\varphi$  be a compactly supported univariate scaling function in  $\text{BV}(\mathbb{R}^1)$  which generates the compactly supported orthogonal wavelet  $\psi$ . For  $d \geq 2$ , we consider the multivariate orthogonal wavelet system  $(\psi_\lambda)_{\lambda \in \Delta}$  obtained from  $\varphi$  and  $\psi$ , and normalized in  $\text{BV}(\mathbb{R}^d)$ . Then the greedy operator*

$$\mathcal{G}_N(f) := \sum_{\lambda \in \Lambda_N(f)} c_\lambda(f) \psi_\lambda,$$

*with  $\Lambda_N(f)$  the set of  $N$  indices  $\lambda \in \Delta$  for which  $|c_\lambda(f)|$  are largest, satisfies*

$$\|f - \mathcal{G}_N(f)\|_{L_{d^*}(\mathbb{R}^d)} + N^{-1/d} |\mathcal{G}_N(f)|_{\text{BV}(\mathbb{R}^d)} \leq C(\varphi, d) K(f, N^{-1/2}; L_{d^*}(\mathbb{R}^d), \text{BV}(\mathbb{R}^d)). \quad (1.2)$$

## 2 The space BV

There are several treatments of the space BV. We mention two valuable references [15, 20] which contain all of the properties of BV functions that we shall need. There are several equivalent definitions of BV. The approach we take below is simply the most direct and convenient for our setting.

Let  $\Omega$  be an open set in  $\mathbb{R}^d$ . We begin with the Sobolev space  $W^1(L_1(\Omega))$  which is the collection of all functions in  $L_1(\Omega)$  such that the distributional gradient  $\nabla f$  is also in  $L_1(\Omega)$ . The semi-norm on this space is

$$|f|_{W^1(L_1(\Omega))} := \|\nabla f\|_{L_1(\Omega)},$$

and the norm for this space is obtained by adding the  $L_1(\Omega)$  norm:

$$\|f\|_{W^1(L_1(\Omega))} := |f|_{W^1(L_1(\Omega))} + \|f\|_{L_1(\Omega)}.$$

The space  $BV(\Omega)$  can now be defined as the set of all  $f \in L_1(\Omega)$  for which there is a sequence  $(f_n)$  satisfying

$$\|f - f_n\|_{L_1(\Omega)} \rightarrow 0, \quad \sup_n |f_n|_{W^1(L_1(\Omega))} < \infty. \quad (2.1)$$

The semi-norm on BV is then defined as

$$\inf_{(f_n)} \liminf_{n \rightarrow \infty} |f_n|_{W^1(L_1(\Omega))}, \quad (2.2)$$

where the infimum is taken over all sequences satisfying (2.1). To see that this definition is equivalent to other definitions of BV the reader should consult Theorems 5.2.1 and 5.3.3 in [20].

We mention a couple of properties of the BV norm that we shall use in this paper.

**Remark 2.1** *In the case  $\Omega = \mathbb{R}^d$ , the functions  $f_n$  appearing in (2.1) and (2.2) can be taken to be in  $C^\infty(\mathbb{R}^d)$  with compact support.*

**Remark 2.2** *Let  $I_0$  be a dyadic cube in  $\mathbb{R}^d$  and  $I_j$ ,  $j = 1, \dots, m$ , be a finite collection of disjoint dyadic cubes each of which is contained in  $I_0$ . Let  $\chi_{I_k}$  be the characteristic function of  $I_k$ ,  $k = 0, \dots, m$ . Then the function  $f = \chi_{I_0} - \sum_{j=1}^m \chi_{I_j}$  has BV semi-norm*

$$|f|_{BV(\mathbb{R}^d)} \leq \sum_{j=0}^m \text{meas}_{d-1}(\partial I_j), \quad (2.3)$$

where  $\partial\Omega$  denotes the boundary of a set  $\Omega \subset \mathbb{R}^d$  and  $\text{meas}_{d-1}$  is the  $(d-1)$  dimensional surface measure.

The second result can be proved directly or derived from the well known co-area formula for BV functions (see [20], p. 231). We have equality in (2.3) if the boundaries of the  $I_j$ ,  $j = 0, 1, \dots, m$ , are disjoint.

**Remark 2.3** *If  $\Omega_j \subset \mathbb{R}^d$ ,  $j = 1, \dots, m$ , is a partition of  $\Omega$ , then*

$$\sum_{j=1}^m |f|_{BV(\Omega_j)} \leq |f|_{BV(\Omega)}. \quad (2.4)$$

This follows from the set additivity of the  $L_1$  semi-norm in the case  $f \in W^1(L_1(\Omega))$  and by taking limits in the general case  $f \in BV(\Omega)$ .

### 3 Wavelet decompositions

We shall limit our analysis to the case of compactly supported orthogonal wavelets on  $\mathbb{R}^d$ . The results we put forward in this paper hold equally well for biorthogonal compactly supported wavelets with the same proofs but somewhat more cumbersome notation.

Let  $\varphi$  be a compactly supported univariate scaling function with orthogonal shifts which satisfies the two scale relation

$$\varphi(x) = \sum_k \alpha_k \varphi(2x - k),$$

where only a finite number of the  $\alpha_k$  are nonzero. We shall assume throughout this paper that  $\varphi$  is in  $BV(\mathbb{R}^1)$ . Let  $\psi$  be the univariate wavelet function with compact support which is obtained from  $\varphi$  by multiresolution. Examples of such wavelets and scaling functions were given by Daubechies [8].

We use the standard construction of multidimensional wavelet bases. Let  $E'$  denote the set of vertices of the cube  $[0, 1]^d$  and  $E$  denote the set of nonzero vertices. We shall use the notation  $\psi^0 := \varphi$  and  $\psi^1 := \psi$ . For each  $e \in E'$ , we define

$$\psi^e(x_1, \dots, x_d) := \psi^{e_1}(x_1) \cdots \psi^{e_d}(x_d).$$

Let  $\mathcal{D}$  denote the set of dyadic cubes in  $\mathbb{R}^d$  and let  $\mathcal{D}_k$  denote those dyadic cubes which have sidelength  $2^{-k}$  and  $\mathcal{D}_+ := \cup_{k \geq 0} \mathcal{D}_k$ . For any dyadic cube  $I = 2^{-k}(j + [0, 1]^d) \in \mathcal{D}_k$ ,  $k \in \mathbb{Z}$ ,  $j \in \mathbb{Z}^d$ , we define the functions

$$\psi_I^e(x) := \gamma(I, e) \psi^e(2^k x - j), e \in E',$$

with the  $\gamma(I, e) > 0$  chosen so that

$$|\psi_I^e|_{BV(\mathbb{R}^d)} = 1, \quad I \in \mathcal{D}, e \in E'.$$

These functions are scaled to  $I$ . It follows that the constants  $\gamma(I, e) = |I|^{-1/d^*} \gamma(e)^1$  with  $d^* := \frac{d}{d-1}$  and therefore we have

$$c_1 \leq \|\psi_I^e\|_{L_{d^*}(\mathbb{R}^d)} \leq c_2, \quad I \in \mathcal{D}, e \in E',$$

with constants  $c_1, c_2$  depending only on  $\varphi$  and  $d$ . In other words, normalization in BV is equivalent to normalization in  $L_{d^*}$ .

To simplify the notation that follows, we introduce the indexing set  $\Delta$  which consists of all pairs  $\lambda = (I, e)$  with  $I \in \mathcal{D}_+$  and  $e \in E$  ( $e \in E'$  if  $I \in \mathcal{D}_0$ ). We define  $|\lambda| := k$  when  $I \in \mathcal{D}_k$ . The set of functions  $\{\psi_\lambda\}_{\lambda \in \Delta}$  is a complete orthogonal system. Any locally integrable function  $f$  on  $\mathbb{R}^d$  has a formal wavelet series

$$f = \sum_{\lambda \in \Delta} c_\lambda(f) \psi_\lambda,$$

where the wavelet coefficients  $c_\lambda(f)$  are given by

$$c_\lambda(f) := c_I^e(f) := \langle f(\cdot), \gamma'(e, I) \psi^e(2^k \cdot - j) \rangle, \quad \lambda = (I, e) \in \Delta, I = 2^{-k}(j + [0, 1]^d),$$

---

<sup>1</sup>Throughout this paper, we shall use the notation  $|A|$  to denote the Lebesgue measure of a set  $A \subset \mathbb{R}^d$ .

where the normalization factors  $\gamma'(e, I)$  of the dual wavelet scale are like  $\gamma'(I, e) \sim |I|^{-1/d}$ .

The set of functions  $\{\psi_\lambda\}_{\lambda \in \Delta}$  is a basis for many function spaces. For example, they are an orthogonal basis for  $L_2(\mathbb{R}^d)$ . They are an unconditional basis for the  $L_p$  spaces  $1 < p < \infty$  and for the Besov spaces whenever they admit an unconditional basis. They are a basis for  $W^1(L_1(\Omega))$ , but not unconditional (this space does not admit an unconditional basis).

We shall use the abbreviated notation  $\phi := \psi^{(0, \dots, 0)}$  for the function which is a tensor product of scaling functions. Similarly, we write

$$\phi_I(x) := |I|^{-\frac{1}{d^*}} \phi(2^k x - j), \quad I = 2^{-k}(j + [0, 1]^d),$$

to index the scaling functions at level  $k$ . The shift invariant space  $\mathcal{S}_k := \mathcal{S}_k(\phi)$  is the span of the functions  $\phi_I$ ,  $I \in \mathcal{D}_k$ . Each space  $\mathcal{S}_k$  is a dilate of the space  $\mathcal{S}_0$ . At each dyadic level  $k$ , the shifts  $\phi_I$ ,  $I \in \mathcal{D}_k$  sum to a constant:

$$\sum_{I \in \mathcal{D}_k} \phi_I = c |I|^{-\frac{1}{d^*}} \quad (3.1)$$

with  $c$  a constant. Any wavelet  $\psi_\lambda$  or scaling function at a dyadic level  $j < k$  (i.e.  $|\lambda| = j$ ) is an element in  $\mathcal{S}_k$  and can be written as a finite linear combination of the  $\phi_I$ ,  $I \in \mathcal{D}_k$ .

## 4 Approximation by piecewise constants

We shall use in the course of our proofs some results on approximation of BV functions by piecewise constant functions. Throughout this and the next section, we assume that  $d \geq 2$ . The results we shall need are for the most part proved in two earlier works [4] (for the case  $d = 2$ ) and [19] (for the case  $d > 2$ ).

We shall discuss three types of approximation by piecewise constants. The first of these is  $N$ -term approximation using Haar functions. In this case, we can be more general and treat  $N$ -term approximation using compactly supported wavelets. So let  $(\psi_\lambda)_{\lambda \in \Delta}$  be one of the wavelet bases introduced in the previous section. We take the basis functions  $\psi_\lambda$  to be normalized in BV.

We define the nonlinear space

$$\Sigma_N^w := \left\{ \sum_{\lambda \in \Lambda} c_\lambda \psi_\lambda : \#(\Lambda) \leq N \right\}.$$

Thus, each element in  $\Sigma_N^w$  is a linear combination of at most  $N$  wavelets which can occur at arbitrary positions or scales.

We define the error in approximating  $f \in L_p(\mathbb{R}^d)$  by the elements of  $\Sigma_N^w$  by

$$\sigma_N^w(f)_{L_p(\mathbb{R}^d)} := \inf_{S \in \Sigma_N^w} \|f - S\|_{L_p(\mathbb{R}^d)}. \quad (4.1)$$

A fundamental result in wavelet approximation [17] is that the approximation error  $\sigma_N^w(f)_{L_p(\mathbb{R}^d)}$  can be obtained up to a constant  $C(d, \varphi)$  by greedy approximation. We describe this result only in the case  $p = d^*$  although it holds for all  $1 < p < \infty$  when

one uses wavelets normalized in  $L_p(\mathbb{R}^d)$ . For each  $N = 1, 2, \dots$ , we define the greedy approximant

$$\mathcal{G}_N(f) := \sum_{\lambda \in \Lambda_N(f)} c_\lambda(f) \psi_\lambda,$$

where  $\Lambda_N(f)$  is the set of the  $N$  indices of the largest coefficients  $c_\lambda(f)$ ,  $\lambda \in \Delta$ , in absolute value (ties in the size of these coefficients can be handled in an arbitrary way). Then, we have

**Proposition 4.1** *For any  $f \in L_{d^*}(\mathbb{R}^d)$ , we have*

$$\|f - \mathcal{G}_N(f)\|_{L_{d^*}(\mathbb{R}^d)} \leq C(d, \varphi) \sigma_N(f)_{L_{d^*}(\mathbb{R}^d)}^w. \quad (4.2)$$

**Proof:** This result can be derived easily from a result of [17] where it is shown that (4.2) holds when  $\mathcal{G}_N(f)$  is replaced by  $\mathcal{G}_N^{L_{d^*}}(f)$ . Here,  $\mathcal{G}_N^{L_{d^*}}(f)$  is defined as above except that one starts with the wavelet coefficients normalized in  $L_{d^*}$  instead of BV.

Let  $\{\psi_\lambda\}_{\lambda \in \Delta}$  be as usual the wavelet basis normalized for BV:  $|\psi_\lambda|_{\text{BV}} = 1$ . For each  $\lambda \in \Delta$ , we choose  $\xi_\lambda$  such that  $\|\xi_\lambda \psi_\lambda\|_{L_{d^*}(\mathbb{R}^d)} = 1$ . The equivalence of the BV and  $L_{d^*}$  normalizations gives that  $c_1 \leq \|\psi_\lambda\|_{L_{d^*}(\mathbb{R}^d)} \leq c_2$ , with  $c_1, c_2 > 0$  independent of  $\lambda$ . Because of the unconditionality of the wavelet basis for  $L_{d^*}(\mathbb{R}^d)$ , there are  $C_1, C_2 > 0$  such that for any sequence of coefficients  $\{a_\lambda\}_{\lambda \in \Delta}$ ,

$$C_1 \left\| \sum_{\lambda \in \Delta} a_\lambda \psi_\lambda \right\|_{L_{d^*}(\mathbb{R}^d)} \leq \left\| \sum_{\lambda \in \Delta} a_\lambda \xi_\lambda \psi_\lambda \right\|_{L_{d^*}(\mathbb{R}^d)} \leq C_2 \left\| \sum_{\lambda \in \Delta} a_\lambda \psi_\lambda \right\|_{L_{d^*}(\mathbb{R}^d)}. \quad (4.3)$$

Given any function  $f = \sum_{\lambda \in \Delta} c_\lambda(f) \psi_\lambda$  in  $L_{d^*}(\mathbb{R}^d)$ , we let  $g := \sum_{\lambda \in \Delta} c_\lambda(f) \xi_\lambda \psi_\lambda$  which by (4.3) is also in  $L_{d^*}(\mathbb{R}^d)$ . If  $\mathcal{G}_N(f) = \sum_{\lambda \in \Delta} c_\lambda(f) \psi_\lambda$  then  $\mathcal{G}_N^{L_{d^*}}(g) = \sum_{\lambda \in \Delta} c_\lambda(f) \xi_\lambda \psi_\lambda$ . Hence, using (4.3), we have

$$C_1 \|f - \mathcal{G}_N(f)\|_{L_{d^*}(\mathbb{R}^d)} \leq \|g - \mathcal{G}_N^{L_{d^*}}(g)\|_{L_{d^*}(\mathbb{R}^d)} \leq C(d, \varphi) \sigma_N^w(g)_{L_{d^*}(\mathbb{R}^d)}. \quad (4.4)$$

On the other hand, if  $S = \sum_{\lambda \in \Delta} a_\lambda \psi_\lambda$  is a best  $N$ -term approximation to  $f$  in  $L_{d^*}(\mathbb{R}^d)$  then, using (4.3) again, we have

$$\sigma_n^w(g)_{L_{d^*}(\mathbb{R}^d)} \leq \|g - \sum_{\lambda \in \Lambda} a_\lambda \xi_\lambda \psi_\lambda\|_{L_{d^*}(\mathbb{R}^d)} \leq C_2 \|f - S\|_{L_{d^*}(\mathbb{R}^d)} = C_2 \sigma_N(f)_{L_{d^*}(\mathbb{R}^d)}. \quad (4.5)$$

The estimates (4.4) and (4.5) combine to prove the proposition.  $\square$

We are interested in quantitative estimates for the approximation error  $\sigma_N(f)_{L_{d^*}(\mathbb{R}^d)}$  whenever  $f \in \text{BV}(\mathbb{R}^d)$ . This will be provided by the following lemma.

**Lemma 4.2** *For any function  $f \in \text{BV}(\mathbb{R}^d)$  we have the estimate*

$$\sigma_N^w(f)_{L_{d^*}(\mathbb{R}^d)} \leq C(d, \varphi) N^{-1/d} |f|_{\text{BV}(\mathbb{R}^d)}, \quad N = 1, 2, \dots$$



**Proof:** The set  $\mathcal{A}_\infty^\alpha(L_p(\mathbb{R}^d))$  of functions  $f \in L_p(\mathbb{R}^d)$  which satisfy

$$\sigma_N^w(f)_{L_p(\mathbb{R}^d)} \leq CN^{-\alpha} \quad (4.6)$$

is called an approximation space. The norm  $\|f\|_{\mathcal{A}_\infty^\alpha(L_p(\mathbb{R}^d))}$  in this space is the smallest  $C > 0$  for which (4.6) is valid. For  $1 < p < \infty$ , and  $\alpha > 0$ , it was proved in [5] that  $f \in \mathcal{A}_\infty^\alpha(L_p(\mathbb{R}^d))$  if and only if the sequence  $(\|(c_\lambda(f)\psi_\lambda)\|_{L_p(\mathbb{R}^d)})_{\lambda \in \Delta}$  is in the space weak  $\ell_\tau$  (denoted by  $w\ell_\tau$ ) with  $\frac{1}{\tau} = \alpha + \frac{1}{p}$ . Moreover,  $\|f\|_{\mathcal{A}_\infty^\alpha(L_p(\mathbb{R}^d))}$  is equivalent to  $\|(\|(c_\lambda(f)\psi_\lambda)\|_{L_p(\mathbb{R}^d)})_{\lambda \in \Delta}\|_{w\ell_\tau}$ . In the case of interest to us, we have  $p = d^* = \frac{d}{d-1}$  and  $\alpha = 1/d$  so that  $\tau = 1$ . It was shown in [4] (for the case of Haar wavelets) and in [6] (for general wavelets) that the wavelet coefficients of a BV function  $f$  are in weak  $\ell_1$  and satisfy

$$\|(\|(c_\lambda(f)\psi_\lambda)\|_{L_{d^*}(\mathbb{R}^d)})_{\lambda \in \Delta}\|_{w\ell_1} \leq C(\varphi, d)|f|_{\text{BV}(\mathbb{R}^d)}.$$

Therefore, the lemma follows.  $\square$

In the case of  $\varphi = \chi_{[0,1]}$ , the wavelets  $\psi_\lambda$ ,  $\lambda \in \Delta$ , are the Haar wavelets and the elements in  $\Sigma_N^w$  are piecewise constant functions which take at most  $CN$  values. We shall now consider two other types of nonlinear approximation using piecewise constants which will be important for us later. For the first of these, let

$$\Sigma_N^c := \left\{ \sum_{I \in \Lambda} c_I \chi_I : \#\Lambda \leq N \right\},$$

where  $\Lambda \subset \mathcal{D}$  is a set of dyadic cubes and for each set  $S$  in  $\mathbb{R}^d$ ,  $\chi_S$  denotes the characteristic function of  $S$ . Note that we do not require that the cubes in  $\Lambda$  are disjoint. In analogy with (4.1), we define

$$\sigma_N^c(f)_{L_p(\mathbb{R}^d)} := \inf_{S \in \Sigma_N^c} \|f - S\|_{L_p(\mathbb{R}^d)}.$$

Since each Haar wavelet  $H_\lambda$  is a linear combination of at most  $2^d$  characteristic functions of dyadic cubes, it follows that  $\Sigma_N^w \subset \Sigma_{2^d N}^c$  and hence from Lemma 4.2, we have

$$\sigma_N^c(f)_{L_{d^*}(\mathbb{R}^d)} \leq C(d, \varphi)N^{-1/d}|f|_{\text{BV}(\mathbb{R}^d)}, \quad N = 1, 2, \dots$$

Lastly, we shall consider approximation by dyadic rings. If  $I$  and  $J \subset I$  are two distinct dyadic cubes ( $J$  maybe the empty set), then we define the dyadic ring  $R = R(I, J)$  to be the set  $R = I \setminus J$ . Consider the nonlinear space

$$\Sigma_N^r := \left\{ \sum_{R \in \mathcal{P}} c_R \chi_R : \#\mathcal{P} \leq N \right\},$$

where  $\mathcal{P}$  is a family of disjoint rings (i.e. any two  $R$  in  $\mathcal{P}$  are disjoint). Note that  $\chi_R = \chi_I - \chi_J$ , and therefore

$$\Sigma_N^r \subset \Sigma_{2N}^c.$$

In analogy with the approximation errors defined above for  $N$ -term approximation by wavelets and constants, we define

$$\sigma_N^r(f)_{L_p(\mathbb{R}^d)} := \inf_{S \in \Sigma_N^r} \|f - S\|_{L_p(\mathbb{R}^d)}.$$

The following lemma concerning approximation by the elements of  $\Sigma_N^r$  was proved in [4] for the case  $d = 2$  and by [19] for the case of general  $d$  in [19] (see Proposition 18).

**Lemma 4.3** *For any function  $f \in \text{BV}(\mathbb{R}^d)$  we have the estimate*

$$\sigma_N^r(f)_{L_{d^*}(\mathbb{R}^d)} \leq C(d, \varphi) N^{-1/d} |f|_{\text{BV}(\mathbb{R}^d)}, \quad N = 1, 2, \dots \quad (4.7)$$

This result was proved in [4, 19] for functions in  $\text{BV}([0, 1]^d)$ . However, we can deduce it for general functions in  $\text{BV}(\mathbb{R}^d)$  using the following argument which we will also apply later in similar settings. First, it is enough to prove this result for functions with compact support since it then follows for general  $f$  by a limiting argument (see the definition of  $\text{BV}(\mathbb{R}^d)$  given in (2.1) and Remark 2.1). Suppose then that  $f$  is supported on  $Q_k := [-2^{k-1}, 2^{k-1}]^d$  for some  $k \geq 1$ . We consider the mapping  $\eta(x) := 2^k(x - e/2)$  where  $e := (1, 1, \dots, 1) \in \mathbb{Z}^d$ . Then  $\eta$ , which is composed of a shift (by  $e/2$ ) and then a dyadic dilation (by  $2^k$ ), maps  $[0, 1]^d$  onto  $Q_k$ . Moreover,  $\eta$  maps any dyadic cube properly contained in  $[0, 1]^d$  into a dyadic cube contained in  $Q_k$ . Now let  $g := f(\eta)$  and apply the analogue of (4.7) for  $[0, 1]^d$  to  $g$ . This result gives a partition  $\mathcal{P}$  with  $\#(\mathcal{P}) \leq N$  and a function  $S = \sum_{R \in \mathcal{P}} c_R \chi_R$ , where the  $R \in \mathcal{P}$  are all of the form  $R = I - J$  with  $I, J$  dyadic subcubes of  $[0, 1]^d$ . If one of these  $R$  has  $I = [0, 1]^d$  then we can replace this  $R$  by at most  $2^d$  rings corresponding to each of the children of  $[0, 1]^d$  and in this way we can assume that any ring in  $\mathcal{P}$  involves dyadic cubes with sidelength  $< 1$ . The function  $S(\eta^{-1})$  is in  $\Sigma_{cN}^r$ . From the fact that  $S$  approximates  $g$  in  $L_{d^*}([0, 1]^d)$  to the accuracy  $C(d, \varphi) N^{-1/d} |g|_{\text{BV}([0, 1]^d)}$  we deduce that  $S(\eta^{-1})$  approximates  $f$  to the accuracy  $C(d, \varphi) N^{-1/d} |f|_{\text{BV}([0, 1]^d)}$  (recall that the  $L_{d^*}$  and  $\text{BV}$  norms scale the same under dilation. This the gives (4.7).

Let us make one last observation about approximation using the elements of  $\Sigma_N^r$ . Given a locally integrable function  $f$ , for each measurable set  $\Omega \subset \mathbb{R}^d$ , we denote by  $f_\Omega$  the average of  $f$  over  $\Omega$ :

$$f_\Omega := \frac{1}{|\Omega|} \int_\Omega f(x) dx.$$

**Lemma 4.4** *If  $f \in \text{BV}(\mathbb{R}^d)$ , there is a collection  $\mathcal{P}$  of disjoint rings  $R$ , such that  $\#(\mathcal{P}) \leq N$  and the function*

$$\mathcal{R}_N(f) := \sum_{R \in \mathcal{P}} f_R \chi_R$$

*satisfies*

$$\|f - \mathcal{R}_N(f)\|_{L_{d^*}(\mathbb{R}^d)} \leq C(\varphi, d) N^{-1/d} |f|_{\text{BV}(\mathbb{R}^d)}. \quad (4.8)$$

**Proof:** Let  $S \in \Sigma_N^r$  satisfy

$$\|f - S\|_{L_{d^*}(\mathbb{R}^d)} \leq C(\varphi, d)N^{-1/d}|f|_{\text{BV}(\mathbb{R}^d)}. \quad (4.9)$$

The existence of such a function is guaranteed by (4.7). We can write  $S = \sum_{R \in \mathcal{P}} c_R \chi_R$ , where  $\mathcal{P}$  is a collection of at most  $N$  disjoint rings. From the disjointness of the rings in  $\mathcal{P}$ , we have

$$\|f - \mathcal{R}_n(f)\|_{L_{d^*}(\mathbb{R}^d)}^{d^*} = \sum_{R \in \mathcal{P}} \|f - f_R\|_{L_{d^*}(R)}^{d^*} + \|f\|_{L_{d^*}(\mathcal{P}^c)}^{d^*}, \quad \mathcal{P}^c := \mathbb{R}^d \setminus (\cup_{R \in \mathcal{P}} R), \quad (4.10)$$

and

$$\|f - S\|_{L_{d^*}(\mathbb{R}^d)}^{d^*} = \sum_{R \in \mathcal{P}} \|f - c_R\|_{L_{d^*}(R)}^{d^*} + \|f\|_{L_{d^*}(\mathcal{P}^c)}^{d^*}. \quad (4.11)$$

On the other hand,

$$\|f - f_R\|_{L_{d^*}(R)} \leq 2 \inf_{c \in \mathbb{R}} \|f - c\|_{L_{d^*}(R)} \leq 2\|f - c_R\|_{L_{d^*}(R)}.$$

This follows from the fact that the mapping  $f \rightarrow f_R$  is a norm one projector on  $L_{d^*}(\mathbb{R}^d)$ . When this is used in (4.10), then (4.11) and (4.9) prove (4.8).  $\square$

## 5 Inverse inequalities

There are certain inequalities (called Bernstein inequalities) which are companion to the Jackson inequalities. It was shown in [4] (for the case  $d = 2$ ) and [19] (for the case  $d > 2$ ) that any  $S \in \Sigma_N^c$  satisfies

$$|S|_{\text{BV}(\mathbb{R}^d)} \leq C(d)N^{1/d}\|S\|_{L_{d^*}(\mathbb{R}^d)}. \quad (5.1)$$

This inequality was proved when  $S$  was supported on  $[0, 1]^d$  in the above references. If  $S \in \Sigma_N^c$ , we can assume that  $\text{supp } S \subset [-K, K]^d$  for some  $K$  and by dilation and shifts we can map  $[-K, K]^d \rightarrow [0, 1]^d$  and deduce the general case (5.1) from that for  $[0, 1]^d$ .

From (5.1), it follows that the same Bernstein inequality holds when  $S \in \Sigma_N^r$  or  $S \in \Sigma_N^w$  when the wavelet is the Haar wavelet. It will follow from the results of this section that the Bernstein inequality also holds for  $\Sigma_N^w$  for general compactly supported wavelets. However, our more general goal is to prove a Bernstein inequality for functions that are a sum of elements from both  $\Sigma_N^w$  and  $\Sigma_N^c$ .

We begin with a local Bernstein inequality between  $\text{BV}(\mathbb{R}^d)$  and  $L_{d^*}(\mathbb{R}^d)$  with  $d^* = \frac{d}{d-1}$ . For any  $I \in \mathcal{D}_k$ , we denote by  $I'$  a general set of the form  $I \setminus \cup_{j=1}^{2^d} J_j$  where each  $J_j$  is a (possibly empty) subcube of the children  $I_j$ ,  $j = 1, \dots, 2^d$ , of  $I$ .

**Lemma 5.1** *For each  $f \in \mathcal{S}_k$  and for each  $I \in \mathcal{D}_k$  and any of the sets  $I'$  we have*

$$|f \chi_{I'}|_{\text{BV}(\mathbb{R}^d)} \leq C(\varphi, d)\|f \chi_{I'}\|_{L_{d^*}(\mathbb{R}^d)}.$$

**Proof:** First of all, by dilation and translation, we can assume  $k = 0$  and that  $I = [0, 1]^d$ . We fix one of the children  $I_j$  of  $I$  and denote by  $I'_j := I_j \setminus J_j$ . It is enough to show

$$|f\chi_{I'_j}|_{\text{BV}(\mathbb{R}^d)} \leq c_0 \|f\chi_{I'_j}\|_{L_1(\mathbb{R}^d)}, \quad j = 1, \dots, 2^d, \quad (5.2)$$

with  $c_0$  a constant depending only on  $\varphi, d$ . Indeed, we have  $\chi_{I'} = \sum_{j=1}^{2^d} \chi_{I'_j}$  and from (5.2)

$$\begin{aligned} |f\chi_{I'}|_{\text{BV}(\mathbb{R}^d)} &\leq \sum_{j=1}^{2^d} |f\chi_{I'_j}|_{\text{BV}(\mathbb{R}^d)} \leq c_0 \sum_{j=1}^{2^d} \|f\chi_{I'_j}\|_{L_1(\mathbb{R}^d)} \\ &= c_0 \|f\chi_{I'}\|_{L_1(\mathbb{R}^d)} \leq c_0 \|f\chi_{I'}\|_{L_{d^*}(\mathbb{R}^d)}, \end{aligned}$$

where the last inequality uses Hölder's inequality and  $|I'| \leq 1$ .

To prove (5.2), we let  $J$  be one of the  $I_j$ , fix  $J$ , and let  $J' = J \setminus J_j$ . The result for the other  $I_j$  will follow by translation. We first observe that since the space  $\mathcal{S}_0$  has dimension  $\leq C(\varphi, d)$  on  $J$ , we have (by equivalence of norms on a finite dimensional space) that

$$\|f\chi_J\|_{L_\infty(\mathbb{R}^d)} \leq c_1 \|f\chi_J\|_{L_1(\mathbb{R}^d)}, \quad f \in \mathcal{S}_0, \quad (5.3)$$

and

$$|f\chi_J|_{\text{BV}(\mathbb{R}^d)} \leq \|f\chi_J\|_{\text{BV}(\mathbb{R}^d)} \leq c_1 \|f\chi_J\|_{L_1(\mathbb{R}^d)}, \quad f \in \mathcal{S}_0, \quad (5.4)$$

with  $c_1$  depending only on  $\varphi$  and  $d$ .

We consider two cases. The first is that  $J'$  is obtained from  $J$  by removing a cube with measure  $\leq \delta$ , where  $\delta$  will be specified in a moment. In this case, we note that

$$\begin{aligned} \|f\chi_{J'}\|_{L_1(\mathbb{R}^d)} &\leq \|f\chi_J\|_{L_1(J')} + |J \setminus J'| \cdot \|f\chi_J\|_{L_\infty(J)} \\ &\leq \|f\chi_J\|_{L_1(J')} + c_1 |J \setminus J'| \cdot \|f\chi_J\|_{L_1(\mathbb{R}^d)}. \end{aligned} \quad (5.5)$$

Now, we select  $\delta := \frac{1}{2c_1}$ . Then, whenever  $|J \setminus J'| \leq \delta$ , we have  $c_1 |J \setminus J'| \leq \frac{1}{2}$ , and therefore (5.5) gives

$$\|f\chi_{J'}\|_{L_1(\mathbb{R}^d)} \leq 2 \|f\chi_{J'}\|_{L_1(\mathbb{R}^d)}. \quad (5.6)$$

Next, we note that

$$\begin{aligned} |f\chi_{J'}|_{\text{BV}(\mathbb{R}^d)} &\leq |f\chi_J|_{\text{BV}(\mathbb{R}^d)} + \text{meas}_{d-1}(J \setminus J') \|f\chi_J\|_{L_\infty(\mathbb{R}^d)} \\ &\leq c_1 (1 + \text{meas}_{d-1}(J \setminus J')) \|f\chi_J\|_{L_1(\mathbb{R}^d)} \\ &\leq c_1 (1 + 2^{d+1}) \|f\chi_J\|_{L_1(\mathbb{R}^d)} \leq 2c_1 (1 + 2^{d+1}) \|f\chi_{J'}\|_{L_1(\mathbb{R}^d)}. \end{aligned}$$

In the above inequalities, we have used relations (5.3), (5.4), (5.6), and the fact that  $\text{meas}_{d-1}(J \setminus J') \leq 2^{d+1}$ . Thus, we have proved (5.2) in the case  $|J \setminus J'| \leq \delta$ .

To complete the proof, we consider the case when the dyadic cube  $J \setminus J'$  has measure  $> \delta$ . For each such  $J'$  we have (by equivalence of norms on the finite dimensional space  $\mathcal{S}_0|_{J'}$ , see for comparison (5.4))

$$\|f\chi_{J'}\|_{\text{BV}(\mathbb{R}^d)} \leq c(J', \varphi, d) \|f\chi_{J'}\|_{L_1(\mathbb{R}^d)}.$$

There is a finite number of such sets  $J'$  and therefore by enlarging the constant from the first case (if necessary), we obtain (5.2) for all  $J'$  in the second case as well. This proves (5.2) and as noted earlier proves the Lemma.  $\square$

We shall utilize a construction given in [10]. Let  $\Lambda$  be any finite collection of dyadic cubes. Given  $I \in \Lambda$ , we define the set  $B(I) = B(I, \Lambda)$  of maximal cubes in  $I$ :

$$B(I, \Lambda) := \{J \in \Lambda : J \subset I, J \neq I \text{ and if } J' \in \Lambda \text{ with } J' \subset I, J' \neq I, J' \cap J \neq \emptyset \text{ then } J' \subseteq J\}.$$

The following Lemma was proved in [10].

**Lemma 5.2** *If  $\Lambda \subset \mathcal{D}$  is any finite collection of dyadic cubes, then there exists a set of dyadic cubes  $\tilde{\Lambda}$  such that*

- (i)  $\Lambda \subset \tilde{\Lambda}$  and  $\#\tilde{\Lambda} \leq 2^d \#\Lambda$ ,
- (ii) For each cube  $I \in \tilde{\Lambda}$ ,  $\#(B(I, \tilde{\Lambda})) \leq 2^d$ , where the  $B(I, \tilde{\Lambda})$  are defined relative to  $\tilde{\Lambda}$ .
- (iii) each child of  $I$  contains at most one cube from  $B(I, \tilde{\Lambda})$ .

Let us note that in [10] this lemma was proved for the case when the cubes in  $\Lambda$  are contained in  $[0, 1]^d$ . However, we can deduce the lemma as stated above from this by using the following reasoning. It follows by shifts of dyadic cubes that the lemma is true if all of the dyadic cubes of  $\Lambda$  are contained in a single dyadic cube of sidelength one. In the general case given in the above lemma, we can by dilating (if necessary) assume that all dyadic cubes in  $\Lambda$  are contained in  $[-1, 1]^d$ . We can then partition  $\Lambda = \cup_{j=1}^{2^d} \Lambda_j$ , where  $\Lambda_j$ ,  $j = 1, \dots, 2^d$ , is the set of cubes in  $\Lambda$  that are contained in  $I_j$ , where  $I_j$  is one of the  $2^d$  dyadic cubes of sidelength one that make up  $[-1, 1]^d$ . We apply the lemma (as stated in [10]) to each  $\Lambda_j$  to receive  $\tilde{\Lambda}_j$ . Then  $\tilde{\Lambda} := \cup_{j=1}^{2^d} \tilde{\Lambda}_j$  satisfies the above lemma.

We introduce one final notation before stating the main result of this section. Given a dyadic cube  $I \in \mathcal{D}$ , let

$$S(I) := \{J \in \mathcal{D} : |J| = |I|, \text{supp } \phi_I \cap J \neq \emptyset\}.$$

The cubes in  $S(I)$  are called the *support cubes* of  $\phi_I$ . It is clear that  $\#(S(I)) \leq C(\varphi, d)$  because  $\varphi$  has compact support.

The following theorem is the main result of this section. It establishes a Bernstein inequality for hybrid linear combinations of scaling functions and characteristic functions of dyadic cubes.

**Theorem 5.3** *If  $\Lambda_1, \Lambda_2 \subset \mathcal{D}$  each has cardinality at most  $N$  (i.e.  $\#\Lambda_1, \#\Lambda_2 \leq N$ ), then any function*

$$f = \sum_{K \in \Lambda_1} a_K \phi_K + \sum_{K \in \Lambda_2} b_K \chi_K, \tag{5.7}$$

*satisfies*

$$\|f\|_{BV(\mathbb{R}^d)} \leq C(\varphi, d) N^{1/d} \|f\|_{L_{d^*}(\mathbb{R}^d)}.$$

**Proof:** By dilating  $f$  (if necessary), we can assume that each of the functions  $\phi_K$  and  $\chi_K$  appearing in (5.7) are supported in  $[-1, 1]^d$ . (Recall again that the BV and  $L_{d^*}$  norms scale the same under dilation.) Let  $I_j$ ,  $j = 1, \dots, 2^d$ , be the dyadic cubes of sidelength one that make up  $[-1, 1]^d$ . We define

$$\Lambda := \left( \bigcup_{K \in \Lambda_1} S(K) \right) \cup \Lambda_2 \cup \{I_j : j = 1, \dots, 2^d\}.$$

We now apply Lemma 5.2 and receive the set  $\tilde{\Lambda}$  with  $\#(\tilde{\Lambda}) \leq C(\varphi, d)N$ .

For each  $I \in \tilde{\Lambda}$  we now define  $I' := I \setminus \bigcup_{J \in B(I)} J$ , where  $B(I) = B(I, \tilde{\Lambda})$ . We have  $\bigcup_{I \in \tilde{\Lambda}} I' = [-1, 1]^d$  and the sets  $I'$  are pairwise disjoint. Therefore

$$f = \sum_{I \in \tilde{\Lambda}} f \chi_{I'}. \quad (5.8)$$

**Claim:** For any  $I \in \tilde{\Lambda}$  with  $I \in \mathcal{D}_k$ , each summand appearing in the representation (5.7) is in  $\mathcal{S}_k$  on  $I'$ .

To prove this claim, we first consider any  $\phi_K$ ,  $K \in \Lambda_1$ , appearing in the first sum. If  $|K| \geq |I|$  then  $\phi_K \in \mathcal{S}_k$  and we have our claim for this term. In the case  $|K| < |I|$  let  $J$  be any support cube of  $\phi_K$  with  $J \cap I \neq \emptyset$ . Then  $|J| = |K|$  and hence  $J$  is contained in one of the cubes of  $B(I)$  and hence  $J \cap I' = \emptyset$ . Thus such a  $\phi_K$  is zero on  $I'$ . Thus we have established our claim for terms appearing in the first summand.

We now consider an arbitrary term  $\chi_K$  appearing in the second summand for which  $K \cap I \neq \emptyset$ . If  $|K| < |I|$  then  $K$  is contained in one of the cubes in  $B(I)$  which in turn means that  $\chi_K$  is zero on  $I'$ . If  $|K| \geq |I|$  then  $\phi_K$  is identically one on  $I'$ . Since the constant functions are in  $\mathcal{S}_k$ , we have proved our claim for the terms in the second summand as well.

We can now complete the proof of the theorem by returning to (5.8). Because of the claim, we can apply Lemma 5.1 to each term in (5.8) and thereby obtain

$$|f|_{BV(\mathbb{R}^d)} \leq \sum_{I \in \tilde{\Lambda}} |f \chi_{I'}|_{BV(\mathbb{R}^d)} \leq C(\varphi, d) \sum_{I \in \tilde{\Lambda}} \|f\|_{L_{d^*}(\mathbb{R}^d)}.$$

On the other hand, from the Hölder inequality,

$$\begin{aligned} \sum_{I \in \tilde{\Lambda}} \|f \chi_{I'}\|_{L_{d^*}(\mathbb{R}^d)} &\leq (\#(\tilde{\Lambda}))^{1/d} \left( \sum_{I \in \tilde{\Lambda}} \|f \chi_{I'}\|_{L_{d^*}(\mathbb{R}^d)}^{d^*} \right)^{1/d^*} \\ &= (\#(\tilde{\Lambda}))^{1/d} \|f\|_{L_{d^*}(\mathbb{R}^d)} \leq C(\varphi, d) N^{1/d} \|f\|_{L_{d^*}(\mathbb{R}^d)}, \end{aligned}$$

because the sets  $I' \in \tilde{\Lambda}$  are disjoint.  $\square$

Theorem 5.3 contains many Bernstein inequalities as a special case. These are summarized in the following Corollary

**Corollary 5.4** *The Bernstein inequality*

$$|f|_{BV(\mathbb{R}^d)} \leq C(\varphi, d) N^{1/d} \|f\|_{L_{d^*}(\mathbb{R}^d)}$$

is valid whenever

- (i)  $f \in \Sigma_N^w$ ,  $f \in \Sigma_N^c$ ,  $f \in \Sigma_N^r$ .
- (ii)  $f \in \Sigma_N^w \oplus \Sigma_N^r$ .

**Proof:** Indeed, in each of these situations  $f$  can be rewritten in the form (5.7) with each of the two sums in (5.7) having at most  $C(\varphi, d)N$  terms.  $\square$

## 6 Proof of Theorem 1.1

and Theorem 1.2 We can now prove Theorem 1.1. Given  $f \in \text{BV}(\mathbb{R}^d)$ , let  $\mathcal{R}_N(f) \in \Sigma_N^r$  be the function in  $\Sigma_N^r$  satisfying Lemma 4.4. We have

$$\begin{aligned} |\mathcal{G}_N(f)|_{\text{BV}(\mathbb{R}^d)} &\leq |\mathcal{G}_N(f) - \mathcal{R}_N(f)|_{\text{BV}(\mathbb{R}^d)} + |\mathcal{R}_N(f)|_{\text{BV}(\mathbb{R}^d)} \\ &\leq CN^{1/d} \|\mathcal{G}_N(f) - \mathcal{R}_N(f)\|_{L_{d^*}(\mathbb{R}^d)} + |\mathcal{R}_N(f)|_{\text{BV}(\mathbb{R}^d)}. \end{aligned} \quad (6.1)$$

In the last inequality we have used (ii) of Corollary 5.4 for the function  $(\mathcal{G}_N(f) - \mathcal{R}_N(f))$ . We estimate now the first term in (6.1) by

$$\begin{aligned} N^{1/d} \|\mathcal{G}_N(f) - \mathcal{R}_N(f)\|_{L_{d^*}(\mathbb{R}^d)} &\leq N^{1/d} \left( \|\mathcal{G}_N(f) - f\|_{L_{d^*}(\mathbb{R}^d)} + \|f - \mathcal{R}_N(f)\|_{L_{d^*}(\mathbb{R}^d)} \right) \\ &\leq C|f|_{\text{BV}(\mathbb{R}^d)}, \end{aligned} \quad (6.2)$$

where in the last inequality we have used (4.2) and Lemma 4.2 to estimate the first term and Lemma 4.4 to estimate the second term. It follows from Corollary 12 of [19] (see also [4] for the case  $d = 2$ ) that

$$|\mathcal{R}_N(f)|_{\text{BV}(\mathbb{R}^d)} \leq C|f|_{\text{BV}(\mathbb{R}^d)}. \quad (6.3)$$

Here we have used our general arguments of dilation and shifts to deduce (6.3). Using the estimates (6.2) and (6.3) in (6.1) gives the desired estimate.  $\square$

Theorem 1.2 can be proved exactly as Theorem 12 in [19].

## 7 Further discussion

We briefly discuss some further issues which will help put our results into perspective

### 7.1 The case $d = 1$

Theorem 1.1 does not hold in the case  $d = 1$ . Consider, for example, the function  $f = \chi_{[0,1/3]}$  which is in  $\text{BV}([0,1])$ . We take the Haar basis  $H_\lambda$ ,  $\lambda \in \Delta$ , normalized in  $\text{BV}([0,1])$ :  $|\mathcal{H}_\lambda|_{\text{BV}([0,1])} = 1$ . This is the same as normalizing this basis in  $L_\infty([0,1])$ . For each dyadic level  $k = 0, 1, \dots$ , there is exactly one Haar coefficient that is nonzero (it corresponds to the dyadic interval  $I \in \mathcal{D}_k$  which contains  $1/3$ ). This coefficient  $c_\lambda(f)$  has absolute value  $1/3$  so that  $c_\lambda(f)H_\lambda(1/3) = \pm 1/3$ . For any given  $N$ , we can take  $N$  of these intervals so that all of the numbers  $c_\lambda(f)H_\lambda(1/3)$  have the same sign. Then, the function  $\mathcal{G}_N(f)$  obtained by retaining exactly these  $N$  terms of the Haar expansion of  $f$  will have  $\text{BV}([0,1])$  norm  $\geq N/3$ . If one wants to avoid the question of choosing arbitrarily in the case of ties, then one can perturb these coefficients slightly.

## 7.2 Quasi-greedy bases

Let  $X$  be a Banach space and  $\{b_\lambda\}_{\lambda \in \Delta}$  be a (Schauder) basis for  $X$  with  $\|b_\lambda\|_X = 1$ , for all  $\lambda \in \Delta$ . Each  $f \in X$  has a unique basis expansion  $f = \sum_{\lambda \in \Delta} c_\lambda(f) b_\lambda$ . We define the greedy approximant  $\mathcal{G}_N(f)$  as before:

$$\mathcal{G}_N(f) := \sum_{\lambda \in \Lambda_N(f)} c_\lambda(f) b_\lambda,$$

where  $\Lambda_N(f)$  is the set of indices corresponding to the  $N$  largest coefficients in absolute value (with ties handled in an arbitrary way).

The basis  $\{b_\lambda\}$  for the space  $X$  is said to be *quasi-greedy* if  $\|f - \mathcal{G}_N(f)\|_X \rightarrow 0$ ,  $N \rightarrow \infty$ . It is known that the Haar basis is not quasi-greedy for  $L_1(\mathbb{R}^d)$  (see [13]). On the other hand, it follows from what we have proved in this paper, that the wavelet bases are quasi-greedy in  $W^1(L_1(\mathbb{R}^d))$ . Indeed, it was proved in [18] that a basis is quasi-greedy for  $X$  if and only if

$$\|\mathcal{G}_N(f)\|_X \leq C \|f\|_X, \quad f \in X,$$

with  $C > 0$  an absolute constant. From Theorem 1.1, we know that the wavelet bases satisfy

$$|\mathcal{G}_N(f)|_{W^1(L_1(\mathbb{R}^d))} \leq C(\varphi, d) |f|_{W^1(L_1(\mathbb{R}^d))}. \quad (7.1)$$

We want to change from semi-norm to norm in (7.1) which we can accomplish as follows. Since the basis  $\{\psi_\lambda\}_{\lambda \in \Delta}$  is normalized in  $BV(\mathbb{R}^d)$ , it follows that

$$\|\psi_\lambda\|_{L_1(\mathbb{R}^d)} \leq C 2^{-k}, \quad |\lambda| = k.$$

Secondly, we have the embedding  $W^1(L_1(\mathbb{R}^d)) \subset B_\infty^1(L_1(\mathbb{R}^d))$  and

$$\|f\|_{B_\infty^1(L_1(\mathbb{R}^d))} \leq C(d) \|f\|_{W^1(L_1(\mathbb{R}^d))},$$

where  $B_\infty^1(L_1(\mathbb{R}^d))$  is the Besov space whose norm is given by

$$\|f\|_{B_\infty^1(L_1(\mathbb{R}^d))} := \sup_{k \geq 0} \sum_{\lambda \in \mathcal{D}_k} |c_\lambda(f)|.$$

Therefore, taking any index set  $\Lambda$  (not necessarily a greedy selection), we have

$$\left\| \sum_{\lambda \in \Lambda} c_\lambda(f) \psi_\lambda \right\|_{L_1(\mathbb{R}^d)} \leq \sum_{k=0}^{\infty} 2^{-k} \sum_{\lambda \in \Lambda \cap \mathcal{D}_k} |c_\lambda(f)| \leq \|f\|_{W^1(L_1(\mathbb{R}^d))}.$$

Hence, we can add the  $L_1(\mathbb{R}^d)$ -norm of  $\mathcal{G}_N(f)$  to the left side of (7.1) and replace the  $W^1(L_1(\mathbb{R}^d))$  semi-norm of  $f$  by the  $W^1(L_1(\mathbb{R}^d))$  norm and obtain that

$$\|\mathcal{G}_N(f)\|_{W^1(L_1(\mathbb{R}^d))} \leq C(\varphi, d) \|f\|_{W^1(L_1(\mathbb{R}^d))}.$$



### 7.3 Thresholding

Continuing with our setting of a wavelet basis  $\{\psi_\lambda\}_{\lambda \in \Delta}$  normalized for  $\text{BV}(\mathbb{R}^d)$ , for each  $\epsilon > 0$ , we define the hard thresholding operator

$$T_\epsilon(f) := \sum_{\lambda \in \Lambda(f, \epsilon)} c_\lambda(f) \psi_\lambda,$$

where  $\Lambda(f, \epsilon) := \{\lambda : |c_\lambda(f)| > \epsilon\}$ . It follows from Theorem 1.1 that this operator is bounded on  $\text{BV}(\mathbb{R}^d)$ :

$$|T_\epsilon(f)|_{\text{BV}(\mathbb{R}^d)} \leq C(\varphi, d) |f|_{\text{BV}(\mathbb{R}^d)}, \quad f \in \text{BV}(\mathbb{R}^d).$$

There is another version of thresholding (called *soft thresholding*) which is preferred in some problems of statistical optimization. To describe soft thresholding, we fix a function  $\eta(t)$  defined on  $[0, \infty)$  such that  $\eta$  is increasing and

$$\begin{aligned} 0 \leq \eta(t) \leq 1 & \quad \text{for all } t, \\ \eta(t) = 0 & \quad \text{for } 0 \leq t \leq 1/2, \\ \eta(t) = 1 & \quad \text{for } t \geq 1. \end{aligned}$$

Given  $\epsilon > 0$  and  $f \in \text{BV}(\mathbb{R}^d)$ , we define the soft thresholding operator  $T_\epsilon^\eta$  by

$$T_\epsilon^\eta(f) := \sum_{\lambda \in \Delta} \eta(|c_\lambda(f)|/\epsilon) c_\lambda(f) \psi_\lambda.$$

**Claim:** For each  $\epsilon > 0$ , we have

$$|T_\epsilon^\eta(f)|_{\text{BV}(\mathbb{R}^d)} \leq C(\varphi, d) |f|_{\text{BV}(\mathbb{R}^d)}, \quad f \in \text{BV}(\mathbb{R}^d).$$

**Proof of Claim:** We order the coefficients  $c_\lambda := c_\lambda(f)$  of  $f$  in decreasing order as  $|c_{\lambda_1}| \geq |c_{\lambda_2}| \geq \dots$ . We fix integers  $N_0 < N_1 < \dots < N_s$  and numbers  $1 =: \beta_0 > \beta_1 > \dots > \beta_s > \beta_{s+1} := 1/2$  in such a way that

$$\begin{aligned} |c_{\lambda_j}| \geq \epsilon = \beta_0 \epsilon & \quad \text{for } j \leq N_0, \\ |c_{\lambda_j}| \leq \epsilon/2 = \beta_{s+1} \epsilon & \quad \text{for } j > N_s, \\ |c_{\lambda_j}| = \beta_{i+1} \epsilon & \quad \text{for } N_i < j \leq N_{i+1}, \quad i = 0, \dots, s-1. \end{aligned}$$

One checks that

$$T_\epsilon^\eta(f) = \sum_{i=0}^s [\eta(\beta_i) - \eta(\beta_{i+1})] \mathcal{G}_{N_i}(f),$$

so from the triangle inequality we get

$$\begin{aligned} |T_\epsilon^\eta(f)|_{\text{BV}(\mathbb{R}^d)} & \leq \sum_{i=0}^s [\eta(\beta_i) - \eta(\beta_{i+1})] |\mathcal{G}_{N_i}(f)|_{\text{BV}(\mathbb{R}^d)} \\ & \leq C(\varphi, d) |f|_{\text{BV}(\mathbb{R}^d)} \sum_{i=0}^s [\eta(\beta_i) - \eta(\beta_{i+1})] \\ & = C(\varphi, d) |f|_{\text{BV}(\mathbb{R}^d)}. \end{aligned}$$

Note that the above claim and its proof, although stated for the space  $\text{BV}(\mathbb{R}^d)$ , hold for any Banach space  $X$  (used in place of  $\text{BV}(\mathbb{R}^d)$ ) which has a quasi-greedy basis (used in place of  $\{\psi_\lambda\}$ ).

## 7.4 The case of domains $\Omega \subset \mathbb{R}^d$

Versions of Theorem 1.1 remain valid for  $BV(\Omega)$  with  $\Omega$  certain domains in  $\mathbb{R}^d$ . We briefly mention two of the typical settings.

For certain domains  $\Omega \subset \mathbb{R}^d$ , one can construct wavelet bases  $\{\psi_\lambda\}_{\lambda \in \Delta}$  such that  $\text{supp}(\psi_\lambda) \subset \Omega$  for each  $\lambda \in \Delta$ . The  $\psi_\lambda$  whose support is sufficiently inside the interior of the domain are the usual wavelets on  $\mathbb{R}^d$ . Near the boundary, the  $\psi_\lambda$  have a different structure. The first examples of such constructions were made in [3] for an interval on  $\mathbb{R}$ . These constructions were then extended to certain multidimensional domains (such as polyhedral domains) (see [7]) and then ultimately to quite general domains in [1]. These constructed bases have the three main properties we need to prove Theorem 1.1. They are of compact support. The scaling functions on a given dyadic level form a partition of unity. The scaling functions and wavelets on a dyadic level  $k$  can be written as a linear combination of a fixed number of scaling functions at level  $k + 1$ . Thus, an analogue of Theorem 1.1 is valid for such basis where now the  $BV(\mathbb{R}^d)$  norm is replaced by the  $BV(\Omega)$  norm.

The second setting applies to quite general domains  $\Omega \subset \mathbb{R}^d$ . For example, it is sufficient that  $\Omega$  is a Lipschitz graph domain (a minimally smooth domain in the sense of Stein (see [16], p.180)). Any function in  $BV(\Omega)$  can be extended to a function  $Ef$  in  $BV(\mathbb{R}^d)$  satisfying

$$\|Ef\|_{BV(\mathbb{R}^d)} \leq C(\Omega)\|f\|_{BV(\Omega)}.$$

Such extension theorems are typically proved for the space  $W^1(L_1(\Omega))$  and then follow for  $BV(\Omega)$  by a limiting argument. We can expand  $Ef$  in a wavelet expansion

$$Ef = \sum_{\lambda \in \Delta} c_\lambda(Ef)\psi_\lambda.$$

This decompositions serves as a wavelet representation for  $f$  on  $\Omega$ :

$$f = \sum_{\lambda \in \Delta(\Omega)} c_\lambda(Ef)\psi_\lambda,$$

where  $\Delta(\Omega)$  is the set of all indices  $\lambda \in \Delta$  for which  $\psi_\lambda$  does not vanish identically on  $\Omega$ .

Consider now the thresholding operator  $T_\epsilon$  applied to  $f$  and  $Ef$ . Since  $T_\epsilon(f) = T_\epsilon(Ef)$  on  $\Omega$ , we deduce that

$$\begin{aligned} |T_\epsilon(f)|_{BV(\Omega)} &= |T_\epsilon(Ef)|_{BV(\Omega)} \leq |T_\epsilon(Ef)|_{BV(\mathbb{R}^d)} \\ &\leq C(\varphi, d)|Ef|_{BV(\mathbb{R}^d)} \leq C(\varphi, d, \Omega)\|f\|_{BV(\Omega)}. \end{aligned}$$

In general, we cannot replace the norm on the right by the semi-norm.

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