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Mathieu Sart. Non linear wavelet density estimation on the real line. 2023. hal-03915562v3

HAL Id: hal-03915562 https://hal.science/hal-03915562v3

Preprint submitted on 2 Mar 2023 (v3), last revised 6 Nov 2023 (v4)

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NON LINEAR WAVELET DENSITY ESTIMATION ON THE REAL LINE

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ABSTRACT. We investigate the problem of density estimation on the real line \mathbb{R} under \mathbb{L}^1 loss. We carry out a new way to select the important coefficients in some wavelet expansions. We study the resulting estimator when the density is smooth with dominated tails. This assumption is very mild and allows in particular to deal with singularities, spatially inhomogeneous smoothness, and fat tailed distributions.

1. INTRODUCTION.

We consider a real valued random variable X and suppose that its distribution admits a density f with respect to the Lebesgue measure. Our aim is to estimate f from the observation of n independent copies X_1, \ldots, X_n of X.

The method we use is based on a decomposition of the density in a (bi-orthogonal) wavelet basis. The challenge lies in the choice of the coefficients to be estimated and those to be set to zero. A successful choice leads to an estimator \hat{f} that is neither over-smoothed nor under-smoothed. A way to check this point mathematically is to consider a class \mathscr{F} of functions and to compute

$$R_{\ell}(\hat{f},\mathscr{F}) = \sup_{f\in\mathscr{F}} \mathbb{E}\left[\ell(f,\hat{f})\right].$$

In this formula, ℓ is a loss function that is up to the statistician, for instance $\ell = d_q^q$ where d_q is the \mathbb{L}^q distance. The above maximal risk can then be compared to the minimax risk

$$R_{\ell}(\mathscr{F}) = \inf_{\tilde{f}} \sup_{f \in \mathscr{F}} \mathbb{E}\left[\ell(f, \tilde{f})\right].$$

Naturally, $R_{\ell}(\mathscr{F}) \leq R_{\ell}(\hat{f}, \mathscr{F})$ and we expect the reverse inequality to be true to within a multiplicative factor. If this factor does not depend on n, \hat{f} is said to be rate optimal (and nearly rate optimal if the factor grows as $\log^k n$).

A wavelet estimator is usually studied under the assumption that f belongs to a ball $\mathfrak{B}_{p,\infty}^{\alpha}(R)$ of a Besov space. The precise definition of this set is recalled in Section 3.1 below. The parameter α indicates the regularity of f whereas p measures, in some sense, the sparsity of its wavelet expansion. This assumption of regularity is seldom the only one. There are two points to watch out for when assessing the quality of a wavelet estimator. First, we have to examine the precise set $\mathscr{F} \subset \mathfrak{B}_{p,\infty}^{\alpha}(R)$ on which it is rate optimal (or at least nearly rate optimal). Second, we have to look at the conditions on α and p. The larger \mathscr{F} is, and the weaker the conditions, the better.

Date: March, 2023.

²⁰¹⁰ Mathematics Subject Classification. 62G05, 62G07.

Key words and phrases. Besov classes, minimax rates, non-parametric estimation, wavelet methods.

The simplest wavelet estimation method is to keep all the coefficients up to a certain resolution and discard the others. This leads to a linear estimator that is – if the final resolution is properly chosen – rate optimal for the \mathbb{L}^q loss when f is a compactly supported density of $\mathfrak{B}_{p,\infty}^{\alpha}(R)$ with $p \ge q$. The case p < q is more delicate and can only be solved with non-linear methods, see [DJKP96]. The two main rules that we know of consist in thresholding the coefficients individually or by block. For more details about them, we refer to [Aut06, CC05, DJ96, DJKP96, HKP98, HKPT12, KPT96]. Note that unwanted logarithmic factors may be involved in the rates of these estimators. Undesirable restrictions on f or on α (such as $\alpha > 1/p$) may also be assumed. We refer to [Sar21] for a description of the minimax rates in the compact case, within log factors, but without superfluous conditions on f or on the parameters.

In the present manuscript, we pay particular attention to the \mathbb{L}^1 loss. The spatial adaptivity of a non-linear estimator may be revealed by studying its risk when the density is compactly supported and in $\mathfrak{B}_{p,\infty}^{\alpha}(R)$ with p < 1. The optimal estimation rate in this context is $n^{-\alpha/(2\alpha+1)}$ for all $\alpha > 1/p - 1$, see [Bir06a]. A wavelet estimator should therefore reach this rate under the sole condition that α belongs to an interval of the form $(1/p - 1, \tau)$ (this restriction on α ensures the classical characterisation of Besov spaces in terms of wavelet coefficients). We have not been able to find in the literature a local or block thresholded wavelet estimator with this property. This will be the case for the estimator we propose in this paper.

The wavelet approach is less developed in the context of density estimation on the real line \mathbb{R} . The main papers we know on this subject are [JLL04, RBRTM11]. In both cases, the estimator is term-by-term thresholded. It is also nearly optimal over some balls $\mathfrak{B}_{p,\infty}^{\alpha}(R)$ of Besov spaces. More precisely, the first paper deals with $p = \infty$ and q > 1 whereas the second one covers the cases $p \ge 1$ and q = 2 (under an additional boundedness assumption when p < 2 though). Other solutions to wavelets have certainly been proposed in the literature to cope with infinite supports. It is beyond the scope of this introduction to describe all the results already obtained. We simply cite here the papers of [GL11, GL14, Lep13, LW19]. They contain the most general results we know about the minimax rates.

Let us mention that difficulties occur when the estimation is performed on \mathbb{R} with the \mathbb{L}^1 loss. This distance gives more weight to the estimation errors in the tails of f than the other \mathbb{L}^q distances. It follows that a pure regularity assumption – such as $f \in \mathcal{B}^{\alpha}_{\infty,\infty}(R)$ with R large enough – is not sufficient to ensure the convergence to 0 of the minimax risk. A way to bypass this problem is to add a constraint on the tails of f, see the seminal paper of [GL14]. This latter paper also contains the following important result: the optimal estimation rate of a (bounded) function in a Besov ball with $p \geq 1$ is, within logarithmic factors, $n^{-\gamma}$ where $\gamma \in [0, \alpha/(2\alpha + 1)]$ depends on the tails of f. If we ignore these log factors, it means that the usual rate $n^{-\alpha/(2\alpha+1)}$ can be recovered when the tail of the distribution is not too fat. We will see below that this result can be refined.

We propose in Section 2 a new estimation method based on wavelets. Each coefficient can, a priori, be retained or discarded at the end of our procedure. However, the decision to keep or delete a coefficient is not purely local. Instead, it is based on a heuristic assessment of the global \mathbb{L}^1 risk of our estimator. The main interest of this reasoning is that it gets rid of undesirable logarithmic factors in almost all cases. Moreover, the resulting estimator is automatically adaptive as our method only uses the data to determine the highest resolution and the coefficients to estimate. No further information is required (such as, for instance, the supremum norm of f or R when $f \in \mathfrak{B}_{p,\infty}^{\alpha}(R)$).

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We study the risk of our estimator under two conditions. The first is that f belongs to a Besov ball $\mathfrak{B}_{p,\infty}^{\alpha}(R)$ with p > 0 and $\alpha \in ((1/p-1)_+, \tau)$ where τ only depends on the choice of the wavelet basis. The second condition concerns the tails of f. We show that our estimator is always rate optimal, except possibly at a limit case, where it is, in the worst case scenario, nearly rate optimal. In particular, we recover the rate $n^{-\alpha/(2\alpha+1)}$ associated with compactly supported densities when the tails of f are not too fat. The assumption $f \in \mathfrak{B}_{p,\infty}^{\alpha}(R)$ alone is also sufficient to ensure the convergence of our estimator when p < 1. This claim contrasts with what happens when $p \ge 1$, where, on the contrary, the minimax risk may not tend to zero. We also investigate the case $\alpha \in (0, 1/p-1]$ and show that the condition $\alpha > (1/p-1)_+$ is, in some sense, necessary to estimate the density (even if it is compactly supported). A precise formulation of these results may be found in Sections 3.2, 3.3 and 3.4.

For ease of presentation, we have only discussed balls $\mathfrak{B}_{p,\infty}^{\alpha}(R)$ of Besov spaces above. However, our bounds are slightly more general. The Besov balls can be replaced in almost all cases by subsets $\mathcal{WB}_{p,\infty}^{\alpha}(R)$ of weak Besov spaces without changing the convergence rate of our estimator. These sets are defined in the same section as the Besov balls, i.e. in Section 3.1.

Finally, let us mention that the computational complexity of our procedure is nearly linear in the number n of observations. We give more details on this point in Section 3.5.

Throughout the paper, we suppose $n \ge 2$ and denote by |A| the size of a finite set A. The letters c, C, c', \ldots stand for quantities that may change from line to line. The proof of our main result is deferred in Section 4. The appendix contains additional proofs as well as some results used in Section 4.

2. Estimation procedure

2.1. Bi-orthogonal wavelet basis. Our estimation method relies on a decomposition of the density in a bi-orthogonal wavelet basis. In such a basis, any square integrable function f takes the form

(1)
$$f = \sum_{k \in \mathbb{Z}} \alpha_k \bar{\phi}_k + \sum_{j=0}^{\infty} \sum_{k \in \mathbb{Z}} \beta_{j,k} \bar{\psi}_{j,k},$$

where for any $j \ge 0$ and $k \in \mathbb{Z}$,

$$\alpha_k = \int f(x)\phi_k(x) \,\mathrm{d}x, \quad \beta_{j,k} = \int f(x)\psi_{j,k}(x) \,\mathrm{d}x,$$

where for any $x \in \mathbb{R}$,

$$\phi_k(x) = \phi(x-k), \quad \psi_{j,k}(x) = 2^{j/2}\psi(2^j x - k),$$

$$\bar{\phi}_k(x) = \bar{\phi}(x-k), \quad \bar{\psi}_{i,k}(x) = 2^{j/2}\bar{\psi}(2^j x - k),$$

and where $\phi, \bar{\phi}, \psi, \bar{\psi}$ stand for dual father and mother wavelets. Equality (1) can be shortened by setting $\beta_{-1,k} = \alpha_k, \psi_{-1,k} = \phi_k, \bar{\psi}_{-1,k} = \bar{\phi}_k$. It then becomes

(2)
$$f = \sum_{j=-1}^{\infty} \sum_{k \in \mathbb{Z}} \beta_{j,k} \bar{\psi}_{j,k}.$$

In the following, we suppose that these four wavelets are bounded and compactly supported. Moreover, we take $\phi = 1_{[0,1]}$, and suppose that ψ is piecewise constant.

The simplest example of bi-orthogonal basis is the Haar basis where $\bar{\phi} = \phi$ and $\bar{\psi} = \psi = 1_{[0,1/2]} - 1_{[1/2,1]}$. This basis is even orthogonal. In general, however, $\bar{\phi}$ and $\bar{\psi}$ are not necessarily piecewise constant and can instead be smooth. We refer to [CDF92] for the construction of such a bi-orthogonal basis where $\bar{\phi}$ and $\bar{\psi}$ are Hölder continuous with exponent τ ($\tau \in \mathbb{N}^*$ is an arbitrary number to be chosen).

Note that the density f we estimate is not necessarily in $\mathbb{L}^2(\mathbb{R})$. Despite this, the series always converges in \mathbb{L}^1 norm and equality (1) remains valid. This result is based on classical techniques in wavelet analysis, see Appendix H.

2.2. Outline. We present in this section the main ideas of our estimation procedure. It will be described in detail in Section 2.3.

We consider a collection $\mathbf{K} = (K_j)_{j \ge -1}$ of subsets of \mathbb{Z} and define the linear wavelet estimator

$$\hat{f}_{\mathbf{K}} = \sum_{j=-1}^{\infty} \sum_{k \in K_j} \widehat{\beta}_{j,k} \bar{\psi}_{j,k}, \quad \text{where} \quad \widehat{\beta}_{j,k} = \frac{1}{n} \sum_{i=1}^{n} \psi_{j,k}(X_i)$$

denotes the empirical version of $\beta_{j,k}$.

The quality of this estimator depends on the choice of the collection \mathbf{K} by the statistician. Ideally, \mathbf{K} should be chosen in such a way that the risk of the estimator is minimal. Observe that

$$\left\| \hat{f}_{\mathbf{K}} - f \right\|_{1} \le c \sum_{j=-1}^{\infty} \left\{ 2^{-j/2} \sum_{k \in \mathbb{Z}} |\beta_{j,k}| + \mathcal{B}_{j}(K_{j}) + \mathcal{E}_{j}(K_{j}) \right\},$$

where c only depends on the wavelet basis, where

$$\mathcal{B}_j(K_j) = -2^{-j/2} \sum_{k \in K_j} |\beta_{j,k}|,$$

and where

$$\mathcal{E}_j(K_j) = 2^{-j/2} \sum_{k \in K_j} |\widehat{\beta}_{j,k} - \beta_{j,k}|$$

represents the error due to the estimation of the coefficients indexed by K_j .

The term $\mathcal{B}_{i}(K_{i})$ can be estimated by

$$\widehat{\mathcal{B}}_j(K_j) = -2^{-j/2} \sum_{k \in K_j} |\widehat{\beta}_{j,k}|.$$

The triangle inequality ensures that

$$|\widehat{\mathcal{B}}_j(K_j) - \mathcal{B}_j(K_j)| \le \mathcal{E}_j(K_j),$$

and hence

(3)
$$\left\| \widehat{f}_{\mathbf{K}} - f \right\|_{1} \leq c \sum_{j=-1}^{\infty} \left\{ 2^{-j/2} \sum_{k \in \mathbb{Z}} |\beta_{j,k}| + \widehat{\mathcal{B}}_{j}(K_{j}) + 2\mathcal{E}_{j}(K_{j}) \right\}$$

The key to minimizing the \mathbb{L}^1 risk (or at least the above upper-bound) is therefore to control the error $\mathcal{E}_j(K_j)$ with high probability. If $\widehat{\mathcal{E}}_j(K_j)$ denotes a known upper-bound of $\mathcal{E}_j(K_j)$, minimizing

(4)
$$\mathcal{B}_j(K_j) + 2\mathcal{E}_j(K_j)$$

among all the subsets K_j of \mathbb{Z} is a possible strategy to minimize the right-hand side of (3). In other words, we may consider a subset \widehat{K}_j of \mathbb{Z} minimizing (4). It contains all the indices k of the coefficients $\widehat{\beta}_{j,k}$ to be kept. We then estimate f by $\widehat{f}_{\widehat{\mathbf{K}}}$ where $\widehat{\mathbf{K}} = (\widehat{K}_j)_{j \ge -1}$.

In the very particular case where $\widehat{\mathcal{E}}_{j}(K_{j})$ takes the form

(5)
$$\widehat{\mathcal{E}}_j(K_j) = 2^{-1-j/2} \sum_{k \in K_j} \widehat{e}_j(k),$$

the procedure amounts to keeping only the coefficients above a given threshold:

$$\widehat{K}_j = \left\{ k \in \mathbb{Z}, \, |\widehat{\beta}_{j,k}| \geq \widehat{e}_j(k) \right\}.$$

We then recover a local estimation method where the decision of threshold $\widehat{\beta}_{j,k}$ is based solely on the value of $\widehat{\beta}_{j,k}$ and $\widehat{e}_j(k)$.

But the interesting point of the above reasoning is that it can be applied to any bound $\widehat{\mathcal{E}}_j(K_j)$, and not only those satisfying (5). This makes it possible to control the error more accurately. The resulting procedure may not be a term-by-term hard thresholding rule though.

2.3. Estimation procedure. The aim of this section is to define our estimator \hat{f} of f. We refer to Sections 3.2 and 3.3 for the study of its statistical properties and to Section 3.5 for informations on more computational aspects.

We sort the sample in increasing order: $X_{(1)} < X_{(2)} < \cdots < X_{(n)}$. We define the smallest integer $\hat{J} \ge 0$ satisfying

$$\min_{1 \le i \le n-1} \left(X_{(i+1)} - X_{(i)} \right) > 2^{1-J} L_{\psi},$$

where $L_{\psi} \geq 1$ is such that supp $\psi \in [-L_{\psi}, L_{\psi}]$. For each resolution $j \in \{-1, \ldots, \hat{J}\}$, we determine a set \hat{K}_j of integers by using an algorithm similar to the one described in the previous section. This yields our estimator

(6)
$$\widehat{f} = \sum_{j=-1}^{\widehat{J}} \sum_{k \in \widehat{K}_j} \widehat{\beta}_{j,k} \overline{\psi}_{j,k}$$

of f.

We define for $j \geq -1$ and $k \in \mathbb{Z}$,

$$\widehat{\sigma}_{j,k}^{2} = \begin{cases} \frac{1}{n} \sum_{i=1}^{n} \phi^{2}(X_{i} - k) & \text{if } j = -1, \\ \frac{1}{n} \sum_{i=1}^{n} \psi^{2}(2^{j}X_{i} - k) & \text{if } j \ge 0. \end{cases}$$

Let then

$$\widehat{\mathbb{Z}}_j = \left\{ k \in \mathbb{Z}, \ \widehat{\sigma}_{j,k}^2 \neq 0 \right\}$$

be the set of integers k for which $\psi_{j,k}(X_i) \neq 0$ for at least one observation.

We define for $\ell_j \in \mathbb{N}$ and $r \ge \ell_j + j + 1$,

$$\widehat{\mathbb{Z}}_{j,r}(\ell_j) = \left\{ k \in \mathbb{Z}, \quad 2^{-r-1} < \widehat{\sigma}_{j,k}^2 \le 2^{-r} \right\}$$
$$\widehat{\mathbb{Z}}_{j,\ell_j+j}(\ell_j) = \left\{ k \in \mathbb{Z}, \quad \widehat{\sigma}_{j,k}^2 > 2^{-\ell_j - j - 1} \right\}.$$

To each value of ℓ_j corresponds therefore the partition of $\widehat{\mathbb{Z}}_j$ defined by

(7)
$$\widehat{\mathbb{Z}}_j = \bigcup_{r \ge \ell_j + j} \widehat{\mathbb{Z}}_{j,r}(\ell_j)$$

We now explain which coefficients to retain among those indexed by $\widehat{\mathbb{Z}}_{j,r}(\ell_j)$. By gathering all these coefficients and by choosing ℓ_j appropriately, we obtain the set \widehat{K}_j .

We introduce some positive numbers $\rho_{-1}, \varsigma_{-1}, \rho_0, \varsigma_0 > 0$ that will be specified later on, and set $\varsigma_j = \varsigma_0, \ \rho_j = \rho_0$ for $j \ge 1$. We define the map $\widehat{\mathcal{E}}_{j,r,\ell_j}(\cdot)$ for $r = \ell_j + j$ and x > 0 by

$$\widehat{\mathcal{E}}_{j,r,\ell_j}(x) = \sqrt{\frac{x}{n}} + x \frac{\log_+(2^{r+1}/x)}{n} + \sqrt{\frac{\log((j+2)(r+1)n)}{n}} + \frac{\log((j+2)(r+1)n)}{n}$$

where $\log_+(x) = \log(e+x)$. When $r \ge \ell_j + j + 1$, $\widehat{\mathcal{E}}_{j,r,\ell_j}(\cdot)$ is rather defined for x > 0 by

$$\widehat{\mathcal{E}}_{j,r,\ell_j}(x) = x \sqrt{\frac{2^{-r} \log_+ \left(\widehat{\lambda}_{j,r}/x\right)}{n} + x \frac{\log_+ \left(\widehat{\lambda}_{j,r}/x\right)}{n}} + \sqrt{x \frac{2^{-r} \log((j+2)(r+1)n)}{n}} + \frac{\log((j+2)(r+1)n)}{n},$$

where

$$\widehat{\lambda}_{j,r} = \begin{cases} \min\left\{ \left(|\widehat{\mathbb{Z}}_j| + 2(2L_{\psi} + 1)(\log(j+2) + 2\log n + 1) \right), 2^{r+1} \right\} & \text{if } 2^r > \varsigma_j n / \log((j+2)n) \\ \min\left\{ |\widehat{\mathbb{Z}}'_{j,r}|, 2^{r+1} \right\} & \text{if } 2^r \le \varsigma_j n / \log((j+2)n) \end{cases}$$

and where

$$\widehat{\mathbb{Z}}'_{j,r} = \{ k \in \mathbb{Z}, \ 2^{-r-2} < \widehat{\sigma}^2_{j,k} \le 2^{-r+1} \}.$$

It is extended by continuity at x = 0.

We define the criterion $\gamma_{j,r,\ell_j}(\cdot)$ for $K_{j,r,\ell_j} \subset \widehat{\mathbb{Z}}_{j,r}(\ell_j)$ by

$$\gamma_{j,r,\ell_j}(K_{j,r,\ell_j}) = -2^{-j/2} \sum_{k \in K_{j,r,\ell_j}} |\widehat{\beta}_{j,k}| + \rho_j \widehat{\mathcal{E}}_{j,r,\ell_j}(|K_{j,r,\ell_j}|).$$

We pick out $\widehat{K}_{j,r,\ell_j} \subset \widehat{\mathbb{Z}}_{j,r}(\ell_j)$ such that

(8)
$$\gamma_{j,r,\ell_j}\left(\widehat{K}_{j,r,\ell_j}\right) = \min_{K_{j,r,\ell_j} \subset \widehat{\mathbb{Z}}_{j,r}(\ell_j)} \gamma_{j,r,\ell_j}\left(K_{j,r,\ell_j}\right).$$

Since $\widehat{\mathbb{Z}}_{j,r}(\ell_j)$ is finite, the set \widehat{K}_{j,r,ℓ_j} does exist. If several sets minimize this criterion, \widehat{K}_{j,r,ℓ_j} denotes any of them.

The value of ℓ_j that we will keep in the following is the smallest value $\hat{\ell}_j$ satisfying

(9)
$$\sum_{r=\hat{\ell}_j+j}^{\infty} \gamma_{j,r,\hat{\ell}_j} \big(\widehat{K}_{j,r,\hat{\ell}_j} \big) = \min_{\ell_j \in \mathbb{N}} \sum_{r=\ell_j+j}^{\infty} \gamma_{j,r,\ell_j} \big(\widehat{K}_{j,r,\ell_j} \big)$$

Such a value does exist (for ℓ_j large enough, the partition given by (7) remains the same).

We finally set

$$\widehat{K}_j = \bigcup_{r=\hat{\ell}_j+j}^{\infty} \widehat{K}_{j,r,\hat{\ell}_j}$$

and define our estimator \hat{f} by (6).

3. Theoretical results.

We study in this section the properties of our estimator f.

3.1. Classes of functions. We begin by introducing classes of functions corresponding to assumptions on the density to be estimated.

For this purpose, the following notations will be convenient. We denote for $p \in (0, +\infty]$ and $x = (x_k)_{k \in \mathbb{Z}}$ the (quasi) ℓ^p norm of x by

$$\|x\|_p = \begin{cases} \left(\sum_{k \in \mathbb{Z}} |x_k|^p\right)^{1/p} & \text{if } p < \infty\\ \sup_{k \in \mathbb{Z}} |x_k| & \text{if } p = \infty. \end{cases}$$

The weak (quasi) ℓ^p norm of x is defined by

$$\|x\|_{p,\infty} = \begin{cases} \sup_{t>0} t \left(\sum_{k \in \mathbb{Z}} 1_{|x_k| \ge t} \right)^{1/p} & \text{if } p < \infty \\ \sup_{k \in \mathbb{Z}} |x_k| & \text{if } p = \infty. \end{cases}$$

We recall that $||x||_{p,\infty} \leq ||x||_p$ but that the converse is not true in general.

3.1.1. Besov classes. The classical Besov spaces $\mathcal{B}_{p,\infty}^{\alpha}$ possess a characterisation in terms of wavelets coefficients. It follows from [DJ97] that $\mathcal{B}_{p,\infty}^{\alpha}$ may be defined when $p \in [0, +\infty]$ and $\alpha \in ((1/p-1)_+, \tau)$ as the set of functions f of $\mathbb{L}^{\max\{p,1\}}(\mathbb{R})$ satisfying $\|f\|_{\mathcal{B}_{p,\infty}^{\alpha}} < \infty$ where

$$\|f\|_{\mathcal{B}^{\alpha}_{p,\infty}} = \sup_{j\geq -1} \left\{ 2^{j(\alpha+1/2-1/p)} \|\beta_{j,\cdot}\|_p \right\}.$$

The value of τ is an integer depending on the wavelet basis. It is equal to 1 for the Haar basis. It stands for the smoothness of $\bar{\phi}$ and $\bar{\psi}$ when the basis is the bi-orthogonal basis of [CDF92] (see Section 2.1).

We define the Besov ball $\mathfrak{B}_{p,\infty}^{\alpha}(R)$ as the set of functions $f \in \mathcal{B}_{p,\infty}^{\alpha}$ satisfying $||f||_{\mathcal{B}_{p,\infty}^{\alpha}} \leq R$. This set is slightly smaller than the set composed of functions $f \in \mathcal{B}_{p,\infty}^{\alpha}$ satisfying

$$\sup_{j\geq 0} \left\{ 2^{j(\alpha+1/2-1/p)} \|\beta_{j,\cdot}\|_p \right\} \leq R.$$

This latter set is denoted by $\mathcal{B}_{p,\infty}^{\alpha}(R)$ and is called a strong Besov class. The adjective "strong" is added to avoid any ambiguity with what is below. The difference between a (strong) Besov class and a Besov ball lies therefore in the starting point of the index j.

We define the weak Besov class $\mathcal{WB}^{\alpha}_{p,\infty}(R)$ when p is finite as the set of functions $f \in \mathbb{L}^1(\mathbb{R})$ satisfying

$$\sup_{j\geq 0} \left\{ 2^{j(\alpha+1/2-1/p)} \|\beta_{j,\cdot}\|_{p,\infty} \right\} \leq R.$$

When p is infinite, we set $\mathcal{WB}^{\alpha}_{\infty,\infty}(R) = \mathcal{B}^{\alpha}_{\infty,\infty}(R)$. The main difference with the strong Besov classes is therefore the use of the (quasi) weak ℓ^p norm in place of the standard ℓ^p (quasi) norm.

3.1.2. Tail dominance condition. We define two sets $\mathcal{T}_{\theta}(M)$ and $\mathcal{WT}_{\theta}(M)$ corresponding to strong and weak conditions on the tails of f.

We set for $j \ge 0$ and $k \in \mathbb{Z}$,

(10)
$$F_{j,k} = \int_{2^{-j}(k-1/2)}^{2^{-j}(k+1/2)} f(x) \, \mathrm{d}x$$

We define for M > 0 and $\theta \in (0, 1)$, the set $\mathcal{T}_{\theta}(M)$ gathering the integrable non-negative functions f satisfying

(11)
$$\sup_{j\geq 0} \left\{ 2^{-j(1-\theta)} \|F_{j,\cdot}\|_{\theta}^{\theta} \right\} \leq M.$$

We say that the "strong tail dominance condition" is met when $f \in \mathcal{T}_{\theta}(M)$. This terminology "tail dominance condition" is directly borrowed from [GL14] although their condition differs a little from this one (see the next section for more details).

Inequality (11) can be softened by replacing the (quasi) norm with its weak version. This leads to the set $WT_{\theta}(M)$ and the "weak tail dominance condition" $f \in WT_{\theta}(M)$. By definition, it contains therefore all the integrable non-negative functions f such that

$$\sup_{j\geq 0} \left\{ 2^{-j(1-\theta)} \|F_{j,\cdot}\|_{\theta,\infty}^{\theta} \right\} \leq M.$$

These two conditions are satisfied when there are not too many mass in the tails of the distribution. The parameters M and θ tune this amount of mass allowed. The larger θ is, the heavier the tails can be.

The above conditions are defined only when $\theta \neq 0$. The limit case $\theta = 0$ corresponds to compactly supported functions. More precisely, we define $\mathcal{T}_0(M) = \mathcal{WT}_0(M)$ as the collection of integrable functions f satisfying

$$\sup_{j\geq 0} \left\{ 2^{-j} \left| \{k \in \mathbb{Z}, F_{j,k} > 0\} \right| \right\} \le M.$$

We show in Appendix G the elementary proposition below.

Proposition 1. The following assertions hold true:

- 1. If M < 1, the set $\mathcal{T}_{\theta}(M)$ does not contain any density.
- 2. Let $p \in (0,1)$, R > 0, $\alpha \in (1/p-1,\tau)$. If $f \in \mathfrak{B}_{p,\infty}^{\alpha}(R)$, then $f \in \mathcal{T}_p(c_1R^p)$ where c_1 only depends on the wavelet basis and α, p . Conversely, if $f \in \mathcal{B}_{p,\infty}^{\alpha}(R) \cap \mathcal{T}_p(R^p)$, then $f \in \mathfrak{B}_{p,\infty}^{\alpha}(c_2R)$ where c_2 only depends on the wavelet basis and α, p .
- 3. If f is a compactly supported density on [-L, L], then f belongs to $\mathcal{T}_0(2L+2)$.
- 4. If f is a density satisfying $f(x) \leq A^b |x|^{-b}$ for all $|x| \geq 1$ and some A > 0, b > 1, then f belongs to $\mathcal{WT}_{1/b}(M)$ with $M = c_3(1+A)$ and some c_3 only depending on b.

3.2. Minimax rates. We evaluate in this section the risk of our estimator f when f is smooth with dominated tails. More precisely, we consider R > 0, $M \ge 1$, $p \in (0, +\infty]$, $\alpha \in ((1/p - 1)_+, \tau)$, $\theta \in [0, 1) \cap [0, p]$, and

$$\mathscr{F}_{\alpha,p,\theta}(R,M) = \begin{cases} \mathcal{WB}_{p,\infty}^{\alpha}(R) \cap \mathcal{WT}_{\theta}(M) & \text{if } p \neq 1 \\ \mathcal{B}_{p,\infty}^{\alpha}(R) \cap \mathcal{WT}_{\theta}(M) & \text{if } p = 1. \end{cases}$$

We then give an upper-bound of the maximal risk of our estimator when f lies in $\mathscr{F}_{\alpha,p,\theta}(R,M)$. It involves the following quantities:

$$\begin{split} \gamma &= \begin{cases} \alpha/(2\alpha+1) & \text{if } \theta \leq \alpha/(2\alpha+1-1/p) \\ \alpha(1-\theta)/(\alpha+1-\theta/p) & \text{if } \theta > \alpha/(2\alpha+1-1/p) \end{cases} \\ \nu_n &= \begin{cases} \log n & \text{if } \theta = \alpha/(2\alpha+1-1/p) \text{ and } p \neq 1 \\ (\log n)^{2\gamma} & \text{if } \theta = \alpha/(2\alpha+1-1/p) \text{ and } p = 1 \\ 1 & \text{otherwise} \end{cases} \\ \beta_1 &= \begin{cases} 1/(2\alpha+1) & \text{if } \theta \leq \alpha/(2\alpha+1-1/p) \\ (1-\theta)/(1+\alpha-\theta/p) & \text{if } \theta > \alpha/(2\alpha+1-1/p) \\ (1-\theta)/((1-\theta)(2\alpha+1)) & \text{if } \theta \leq \alpha/(2\alpha+1-1/p) \\ (\alpha+1-1/p)/(\alpha+1-\theta/p) & \text{if } \theta > \alpha/(2\alpha+1-1/p). \end{cases} \end{split}$$

Our main result is proved in Section 4 and is as follows:

Theorem 2. Let $p \in (0, \infty]$, $\alpha \in ((1/p - 1)_+, \tau)$, $\theta \in [0, 1) \cap [0, p]$, R > 0, $M \ge 1$. Then, there exist $\bar{\varrho}_{-1}, \varsigma_{-1}, \bar{\varrho}_0, \varsigma_0$ and n_0 such that for all $n \ge n_0$, $\rho_{-1} \ge \bar{\varrho}_{-1}$, $\rho_0 \ge \bar{\varrho}_0$, our estimator \hat{f} defined in Section 2.3 satisfies

(12)
$$\sup_{f\in\mathscr{F}_{\alpha,p,\theta}(R,M)} \mathbb{E}\left[d_1(f,\hat{f})\right] \le c \left[R^{\beta_1} M^{\beta_2} + M \mathbf{1}_{\gamma=1-\theta}\right] \nu_n n^{-\gamma}.$$

Moreover, $\bar{\varrho}_{-1}, \varsigma_{-1}$ are universal and $\bar{\varrho}_0, \varsigma_0$ only depends on ψ . The term c only depends on $\alpha, p, \bar{\phi}, \psi, \bar{\psi}, \theta, \rho_{-1}, \rho_0$, and n_0 only depends on p, α, θ, R, M .

We would like to highlight that the construction of our estimator \hat{f} does not involve the parameters α, p, θ, R, M of the class $\mathscr{F}_{\alpha,p,\theta}(R, M)$. They can therefore be unknown.

For pedagogical reasons, let us consider the case where f is compactly supported in [-L, L] with $L \ge 1$. Proposition 1 entails that $f \in \mathscr{F}_{\alpha,p,0}(R, 4L)$. In particular, there is n_0 only depending on R, M, L, α, p such that for all $n \ge n_0$,

$$\mathbb{E}\left[d_1(f,\hat{f})\right] \le c' R^{1/(2\alpha+1)} L^{(\alpha+1-1/p)/(2\alpha+1)} n^{-\alpha/(2\alpha+1)}.$$

This is the classical estimation rate. It is however attained here under very mild conditions on pand α . This result essentially says that our estimator adapts to local variations of the density (the algorithm can increase the number of coefficients $\beta_{j,k}$ to estimate at the locations where the function varies a lot and on the contrary decrease it at the locations where the function is more flat). Note that there is no requirement that the density be bounded or even in $\mathbb{L}^2(\mathbb{R})$. We do not know of any wavelet estimator that achieves the standard rate $n^{-\alpha/(2\alpha+1)}$ under our assumptions.

In the non compact case, the usual rate of convergence $n^{-\alpha/(2\alpha+1)}$ applies when θ is sufficiently small, that is when the (weak) tail dominance condition is stringent enough. When θ is larger,

however, the estimation rate deteriorates. It becomes particularly slow when θ comes close to 1. We recall that is not possible to estimate a density on \mathbb{R} under the sole assumption that f belongs to a ball $\mathfrak{B}_{p,\infty}^{\alpha}(R)$ of a Besov space with $p \geq 1$ and R large enough when the loss is the \mathbb{L}^1 norm (see [GL14, IK81, JLL04]).

However, the assumption "f belongs to a ball of a Besov space" is sufficient to ensure the convergence of our estimator when p < 1 and $\alpha > 1/p - 1$. Indeed, our tail dominance condition is fulfilled in this case with $\theta = p$. The above theorem can therefore be applied. This gives: for all $p \in (0,1), \alpha \in (1/p - 1, \tau), R \geq 1$, and n large enough,

(13)
$$\sup_{f \in \mathfrak{B}_{p,\infty}^{\alpha}(R)} \mathbb{E}\left[d_1(f,\hat{f})\right] \le c'' R^{\beta_3} \nu_n n^{-\gamma},$$

where

$$\begin{split} \gamma &= \begin{cases} \alpha/(2\alpha+1) & \text{if } p \in (0,1/2] \text{ or } \alpha < (1-p)/(2p-1) \\ 1-p & \text{if } p \in (1/2,1) \text{ and } \alpha \ge (1-p)/(2p-1) \\ \nu_n &= \begin{cases} \log n & \text{if } p \in (1/2,1) \text{ and } \alpha = (1-p)/(2p-1) \\ 1 & \text{otherwise} \end{cases} \\ \beta_3 &= \begin{cases} \alpha/((2\alpha+1)(1/p-1)) & \text{if } p \in (0,1/2] \text{ or } \alpha < (1-p)/(2p-1) \\ p & \text{if } p \in (1/2,1) \text{ and } \alpha \ge (1-p)/(2p-1) \end{cases} \end{split}$$

and where c'' only depends on $\alpha, p, \psi, \bar{\psi}, \bar{\phi}, \rho_{-1}, \rho_0$.

As far as we know, only the case $p \ge 1$ has been studied in the literature of density estimation with infinite support under \mathbb{L}^1 loss. The only papers we are aware of that deal with the subject are the two mentioned below.

First, the authors of [GL14] proposed to estimate the density pointwise. The global risk is then obtained by integrating the pointwise risk. This reasoning has the merit of not depending on a particular loss and of leading to results for all \mathbb{L}^q losses. The downside is that it may lead to undesirable logarithmic factors in the convergence rates. We do not have any here, except at the boundary. Their tail dominance condition is more or less the same as our strong condition. Rigorously, it is at least as stringent as our strong condition (see Proposition 1 of [CL20]). We do not know whether it is equivalent. Note that they also impose a condition on the supremum norm of f. Besides, they restricted themselves to balls of Besov spaces and were not interested in weak Besov classes.

Second, wavelets were used in [CL20] to estimate the density under the same tail condition as our strong condition. Unfortunately, the convergence rate of their estimator is slower than ours and is hence not optimal (their exponent in n is, in absolute value, smaller than ours as soon as $\theta \neq 0$).

Although our result is stated for the \mathbb{L}^1 loss, it can easily be checked that (12) remains true for the distance induced by the Besov norm $\|\cdot\|_{\mathcal{B}^0_{1,1}}$ defined by

$$\|f\|_{\mathcal{B}^{0}_{1,1}} = \sum_{j \ge -1} 2^{-j/2} \|\beta_{j,\cdot}\|_{1}.$$

The proposition below shows not only that our rate is optimal (except possibly at the boundary), but also that the dependency in R and M is the right one. It completes the minimax lower bound of [GL14] and is proved in Appendix E.

Proposition 3. Let $p \in (0, \infty]$, $\alpha \in ((1/p-1)_+, \tau)$ and $\theta \in (0, 1) \cap [0, p]$. There exist some R_0, M_0 such that: for all $R \ge R_0$, $M \ge M_0$ and all n large enough,

$$\inf_{\tilde{f}} \sup_{f \in \mathcal{B}^{\alpha}_{p,\infty}(R) \cap \mathcal{T}_{\theta}(M)} \mathbb{E}\left[d_1(f,\tilde{f})\right] \ge c \left[R^{\beta_1} M^{\beta_2} + M \mathbf{1}_{\gamma=1-\theta}\right] n^{-\gamma},$$

where c only depends on $\alpha, p, \theta, \psi, \bar{\psi}, \bar{\phi}$. Moreover, R_0, M_0 only depend on $\alpha, p, \theta, \psi, \bar{\psi}, \bar{\phi}$.

3.3. Minimax rates over a class of fat tailed distributions. We illustrate here the interest of the weak tail dominance condition compared to its strong version. We consider $A \ge 1$, $\alpha \in ((1/p-1)_+, \tau)$, $p \in [0, \infty]$, b > 1, $b \ge 1/p$, and the class

$$\mathcal{D}(\alpha, p, R, A, b) = \left\{ f \in \mathcal{B}^{\alpha}_{p, \infty}(R), \text{ such that } f(x) \le A^{b} |x|^{-b} \text{ for all } |x| \ge 1 \right\}.$$

Proposition 1 says that the weak tail dominance condition is met with $\theta = 1/b$ and M = c(1 + A).

We deduce from Theorem 2 that for n large enough,

$$\sup_{f \in \mathcal{D}(\alpha, p, R, A, b)} \mathbb{E}\left[d_1(f, \hat{f})\right] \le C\nu_n n^{-\gamma},$$

where

$$\begin{split} \gamma &= \begin{cases} \alpha/(2\alpha+1) & \text{if } b \geq 2 + (1-1/p)/\alpha \\ \alpha(1-1/b)/(\alpha+1-1/(pb)) & \text{if } b < 2 + (1-1/p)/\alpha \\ \end{cases} \\ \nu_n &= \begin{cases} \log n & \text{if } b = 2 + (1-1/p)/\alpha \text{ and } p \neq 1 \\ (\log n)^{2\gamma} & \text{if } b = 2 + (1-1/p)/\alpha \text{ and } p = 1 \\ 1 & \text{if } b \neq 2 + (1-1/p)/\alpha \end{cases} \end{split}$$

and where C only depends on $A, R, \alpha, p, b, \psi, \overline{\psi}, \overline{\phi}, \rho_{-1}, \rho_0$.

The parameter γ , which governs the estimation rate of our estimator, depends on α , b and p. When b is sufficiently large, we recover the usual rate of convergence. The rate is otherwise slower but still minimax (at least when A, R are large enough, and to within log factors in the two limiting cases). This statement can be checked by reviewing the proof of Proposition 3 above.

Let us now observe that the strong tail dominance condition is not fulfilled for $\theta = 1/b$ (whatever M). We rather have $\mathcal{D}(\alpha, p, R, A, b) \subset \mathcal{T}_{\theta}(M_{\theta})$ for all $\theta > 1/b$ and M_{θ} depending on θ (and b, A). If the theorem were shown only for the strong condition, we could apply it only with values of θ larger than 1/b. If done correctly, it gives the right rate of convergence when $b > 2+(1-1/p)/\alpha$. However, this causes problems when $b \leq 2+(1-1/p)/\alpha$ as the exponent then depends on θ . Using a value of θ larger than 1/b leads to a slower convergence rate.

Note that a bounded and unimodal density f belongs to $\mathcal{B}_{1,\infty}^1(R)$ for some R depending on $||f||_{\infty}$ and the wavelet basis only. Such a density belongs therefore to $\mathcal{D}(1, 1, R, A, b)$ if it satisfies $f(x) \leq Ax^{-b}$ for all $|x| \geq 1$. We deduce that \hat{f} converges to f at the rate $n^{-1/3}$ when b > 2. We thus recover the optimal estimation rate of a bounded unimodal density with compact support although f can be infinitely supported here.

3.4. About the condition $\alpha > 1/p-1$ when p < 1. In the previous sections, we always assumed that $\alpha > (1/p-1)_+$. This condition is empty if $p \ge 1$ but not otherwise. Note that it is used in the result of [DJ97] to characterize the Besov balls in terms of wavelet coefficients. To explore what happens in the opposite case, we need therefore to redefine these balls.

We consider some $\alpha, p > 0$ and an arbitrary integer r larger than α . We set for t, h > 0,

$$\Delta_h^r f(t) = \sum_{k=0}^r \binom{r}{k} (-1)^{r-k} f(t+kh)$$

We define the modulus of smoothness

$$\omega_p(f, x) = \sup_{0 < h \le x} \left[\int \left| \Delta_h^r f(t) \right|^p \, \mathrm{d}t \right]^{1/p},$$

and define the Besov ball $\mathfrak{B}_{p,\infty}^{\alpha}(R)$ as the collection of functions $f \in \mathbb{L}^p$ such that

$$||f||_{\mathcal{B}^{\alpha}_{p,\infty}} = ||f||_p + \sup_{x>0} x^{-\alpha} \omega_p(f,x),$$

is not larger than R.

The above (quasi) norm depends a priori on the choice of r. Changing r leads however to an equivalent (quasi) norm, see [DL93]. It is also equivalent to the (quasi) norm that we defined in Section 3.1.1 when $\alpha \in ((1/p-1)_+, \tau)$. To avoid adding unnecessary notations, we have used the same symbol to designate the Besov ball and the (quasi) norm. However, since the (quasi) norms are not equal, but equivalent, there is a slight ambiguity about what $\|\cdot\|_{\mathcal{B}^{\alpha}_{p,\infty}}$ and R are. This has no impact on our results though.

The proposition below shows that it is not possible to obtain a convergent estimator for the \mathbb{L}^1 loss under the sole assumption that f is a compactly supported density of $\mathfrak{B}_{p,\infty}^{\alpha}(R)$ when $\alpha \leq 1/p - 1$. It is proved in Appendix F.

Proposition 4. Let $p \in (0, 1)$, $\alpha \in (0, 1/p - 1]$ and R > 0. Then,

$$\inf_{\tilde{f}} \sup_{\substack{f \in \mathfrak{B}^{\alpha}_{p,\infty}(R)\\supp \ f \subseteq [0,1]}} \mathbb{E} \left[d_1(f,\tilde{f}) \right] \ge 1/4.$$

This result can be compared with what exists in the literature of estimation of a compactly supported density under the \mathbb{L}^q loss with q > 1. The minimax risk does not tend to 0 either when $\alpha = 1/p - 1/q$, see [Sar21]. However, the optimal estimation rate can be made arbitrarily slow by choosing α very close to 1/p - 1/q when q > 1. This phenomenon does not occur here, since the optimal estimation rate is $n^{-\alpha/(2\alpha+1)}$, whatever p > 0 and $\alpha > 1/p - 1$.

3.5. Computational complexity. An estimator is not always derived from a computationally tractable procedure. For example, we have not been able to find in the literature a computationally tractable algorithm that would lead to an optimal estimator in the minimax \mathbb{L}^1 sense when f is compactly supported on [0,1] and in $\mathfrak{B}_{p,\infty}^{\alpha}(R)$ with p < 1. There are admittedly computationally acceptable solutions in the literature to deal with this case, but they seem to be optimal only to within log factors, see [Sar21]. We recall that the case p < 1 is very different from the case $p \ge 1$. The latter is straightforward to solve in the compact case as a simple linear estimator works.

Let us also mention that there are computational difficulties with Kernel estimators. Indeed, the bandwidth should vary with the location so that the estimator adapts to the inhomogeneous smoothness of the density. In this context, procedures that lead to rate optimal estimators have been developed by [Lep15] in the Gaussian white noise model. His solution is based on an algorithm that seems quite difficult to implement though (see his Problem II Section 4.2). Things are a bit simpler when we allow the estimator to be rate optimal within log factors (see [GL14, LW19]).

We show below that the computational complexity of our procedure is nearly linear in the number n of observations.

We consider some $j \in \{-1, \ldots, \hat{J}\}$ and assume that the wavelets functions $\bar{\phi}, \bar{\psi}$ and ψ have been preprocessed. We put $\psi_j = \psi$ if $j \ge 0$ and $\psi_{-1} = \phi$.

We begin by sorting the observations in increasing order $X_{(1)} < \cdots < X_{(n)}$. We remark that $\psi_j(\mathbb{R})$ is finite as ψ_j is piecewise constant. We consider some $y \in \psi_j(\mathbb{R}) \setminus \{0\}$. Then, we find for each $i \in \{1, \ldots, n\}$, the few integers k such that $\psi_j(2^j X_i - k) = y$. We gather all these elements into a vector of size $\mathcal{O}(n)$. By counting in this vector the number of repetitions, we determine the family $(\hat{\beta}_{j,k}(y))_{k \in \mathbb{Z}}$ where

$$\widehat{\beta}_{j,k}(y) = \frac{y}{n} \sum_{i=1}^{n} 1_{\psi_j(2^j X_i - k) = y}.$$

In particular, we deduce

$$\widehat{\beta}_{j,k} = \sum_{y \in \psi_j(\mathbb{R}) \setminus \{0\}} \widehat{\beta}_{j,k}(y)$$

The reasoning is the same for obtaining $(\hat{\sigma}_{j,k}^2)_{k\in\mathbb{Z}}$. So far, the number of calculations performed is at most $\mathcal{O}(n\log n)$.

We consider some $\ell_j \in \mathbb{N}$, $r \geq \ell_j + j$, find the indices in $\widehat{\mathbb{Z}}_{j,r}(\ell_j)$ and the size of $\widehat{\mathbb{Z}}'_{j,r}$. We sort the wavelet coefficients $(\widehat{\beta}_{j,k})_{k\in\widehat{\mathbb{Z}}_{j,r}(\ell_j)}$ in descending order of importance: $|\widehat{\beta}_{j,[1]}| \geq |\widehat{\beta}_{j,[2]}| \geq |\widehat{\beta}_{j,[3]}| \dots$ Finding $\widehat{K}_{j,r,\ell_j} \subset \widehat{\mathbb{Z}}_{j,r}(\ell_j)$ that minimizes (8) amounts to selecting the *s* most important coefficients where *s* minimizes

$$-2^{-j/2}\sum_{k=1}^{s} |\widehat{\beta}_{j,[k]}| + \rho_j \widehat{\mathcal{E}}_{j,r,\ell_j}(s).$$

This set can therefore be built in $\mathcal{O}(n+|\widehat{\mathbb{Z}}_{j,r}(\ell_j)|\log|\widehat{\mathbb{Z}}_{j,r}(\ell_j)|)$ elementary operations.

Note that the number of r to consider is at most $\mathcal{O}(\log n)$. Moreover, the values of ℓ_j of interest are those between 0 and $\mathcal{O}(\log n)$ since the partition given by (7) remains the same when ℓ_j is higher. Since $\sum_{r \geq \ell_j + j} |\widehat{\mathbb{Z}}_{j,r}(\ell_j)| = \mathcal{O}(n)$, we deduce that all the sets \widehat{K}_{j,r,ℓ_j} (when $r \geq \ell_j + j$, $\ell_j \in \mathbb{N}, j \in \{1, \ldots, \widehat{J}\}$ vary) can be obtained in at most $\mathcal{O}((\widehat{J} + 1)n \log^2 n)$ operations. The computation of $\widehat{\ell}_j$ is fast as it requires less than $\mathcal{O}(\log^2 n)$ additional operations. To sum up, $\mathcal{O}((\widehat{J} + 1)n \log^2 n)$ is the maximal number of operations needed to find all the selected coefficients, that is to find the sets $(\widehat{K}_j)_{-1 < j < \widehat{J}}$.

The computational complexity of our procedure is random but can be bounded from above either with high probability or in expectation. We only present the second possibility. We deduce from Lemma 17 page 24 that $\mathbb{E}[\hat{J}] \leq \mathcal{O}(\log n)$ when f belongs to \mathbb{L}^q for some q > 1. This assumption

is fulfilled when $f \in \mathscr{F}_{\alpha,p,\theta}(R,M)$ (see Lemma 25 page 32 if needed). More precisely, we have for all $p \in (0,\infty]$, $\alpha \in ((1/p-1)_+, \tau)$, $\theta \in [0,1) \cap [0,p]$, R > 0, $M \ge 1$,

$$\sup_{f \in \mathscr{F}_{\alpha,p,\theta}(R,M)} \mathbb{E}\left[J\right] \le \mathcal{O}(\log n).$$

In average, the computational complexity is therefore at most $\mathcal{O}(n \log^3 n)$.

4. Proof of Theorem 2.

We begin by carrying out and proving a non-asymptotic risk bound.

Theorem 5. Suppose that $f \in \mathbb{L}^q$ for some q > 1. Then, there exist universal constants $\bar{\varrho}_{-1}, \varsigma_{-1}$ and terms $\bar{\varrho}_0, \varsigma_0$ depending on ψ only such that if $\rho_{-1} \geq \bar{\varrho}_{-1}$ and $\rho_0 \geq \bar{\varrho}_0$, the estimator \hat{f} defined in Section 2.3 satisfies

$$\mathbb{E}\left[\|f-\hat{f}\|_{1}\right] \leq c_{1}\mathbb{E}\left[\sum_{j=-1}^{\widehat{J}} \inf_{\ell_{j}\in\mathbb{N}} \sum_{r=\ell_{j}+j}^{\infty} \inf_{K_{j,r,\ell_{j}}\subset\widehat{\mathbb{Z}}_{j,r}(\ell_{j})\cap\overline{\mathbb{Z}}_{j}} \left\{\widehat{B}_{j,r,\ell_{j}}(K_{j,r,\ell_{j}}) + \rho_{j}\widehat{\mathcal{E}}_{j,r,\ell_{j}}(|K_{j,r,\ell_{j}}|)\right\} 1_{\mathcal{A}}\right] \\$$
(14)
$$+ c_{1}T_{1} + c_{1}T_{2} + c_{2}\frac{\log n}{n^{2}},$$

where

(15)
$$\overline{\mathbb{Z}}_j = \{k \in \mathbb{Z}, f_{j,k} \ge 1/n\},\$$

(16)
$$f_{j,k} = \int f(x) \mathbf{1}_{supp \ \psi_{j,k}}(x) \, \mathrm{d}x,$$

(17)
$$\widehat{B}_{j,r,\ell_j}(K_{j,r,\ell_j}) = 2^{-j/2} \sum_{k \in \left(\widehat{\mathbb{Z}}_{j,r}(\ell_j) \cap \overline{\mathbb{Z}}_j\right) \setminus K_{j,r,\ell_j}} |\beta_{j,k}|,$$

(18)
$$T_1 = \sum_{j=-1}^{\infty} 2^{-j/2} \sum_{k \notin \overline{\mathbb{Z}}_j} |\beta_{j,k}|,$$

(19)
$$T_2 = n \sum_{j=-1}^{\infty} 2^{-j/2} \sum_{k \in \overline{\mathbb{Z}}_j} |\beta_{j,k}| f_{j,k} (1 - f_{j,k})^{n-1},$$

where c_1 only depends on $\psi, \bar{\psi}, \bar{\phi}$, and where c_2 only depends on $\psi, \bar{\psi}, \bar{\phi}, q, ||f||_q$. Moreover, \mathcal{A} is an event on which $\widehat{J} \leq c_0 \log n$ for some c_0 only depending on $q, ||f||_q$, and on which

(20)
$$|\widehat{\mathbb{Z}}_j| \le 2\mathbb{E}[|\widehat{\mathbb{Z}}_j|] + (7/3)(2L_{\psi} + 1)(\log(j+2) + 2\log n + 1)$$

for all $j \geq -1$. Furthermore, on \mathcal{A} we have for all $j \geq -1$, $k \in \mathbb{Z}$,

(21)
$$\sigma_{j,k}^2 \le 2\hat{\sigma}_{j,k}^2 + c_3 \log((j+2)n)/n$$

where $\sigma_{j,k}^2 = \mathbb{E}[\widehat{\sigma}_{j,k}^2]$ and where c_3 only depends on ψ . We also have on \mathcal{A} : (22) $\widehat{\mathbb{Z}}_{-}(\ell_{-}) \subset \mathbb{Z}_{-} \subset \mathbb{Z}_{-}'$

$$\mathbb{Z}_{j,r}(\ell_j) \subset \mathbb{Z}_{j,r} \subset \mathbb{Z}_{j,r} \subset \mathbb{Z}_{j,r}.$$

This embedding is valid for all $j \ge -1$, $\ell_j \in \mathbb{N}$, $r \ge \ell_j + j + 1$ such that

$$2^r \leq \varsigma_j \frac{n}{\log((j+2)n)},$$

and $\breve{\mathbb{Z}}_{j,r}$, $\breve{\mathbb{Z}}'_{j,r}$ are defined by

(23)
$$\breve{\mathbb{Z}}_{j,r} = \left\{ k \in \mathbb{Z}, \quad 2^{-r-3/2} < \sigma_{j,k}^2 \le 2^{-r+1/2} \right\},$$

(24)
$$\breve{\mathbb{Z}}'_{j,r} = \left\{ k \in \mathbb{Z}, \quad 2^{-r-5/2} < \sigma_{j,k}^2 \le 2^{-r+3/2} \right\}$$

By convention, the infimum over K_{j,r,ℓ_j} in (14) is equal to 0 if $\widehat{\mathbb{Z}}_{j,r}(\ell_j) \cap \overline{\mathbb{Z}}_j = \emptyset$.

4.1. Proof of Theorem 5. For all $j \ge -1$, subset $K \subset \mathbb{Z}$, we define

$$\widehat{\sigma}_{j}^{2}(K) = \sum_{k \in K} \widehat{\sigma}_{j,k}^{2}$$

and

(25)
$$\sigma_j^2(K) = \mathbb{E}\left[\widehat{\sigma}_j^2(K)\right] = \sum_{k \in K} \sigma_{j,k}^2.$$

The proof of the theorem ensues from a succession of lemmas. The first one is classical and is the following:

Lemma 1. For all $K \subset \mathbb{Z}$, $j \ge 0$, and $x \in \mathbb{R}$,

$$\left(\sum_{k\in K} |\psi(2^j x - k)|\right)^2 \le c \sum_{k\in K} \psi^2(2^j x - k),$$

where c only depends on ψ .

Proof of Lemma 1. We use that ψ is compactly supported and apply Cauchy-Schwarz inequality.

Lemma 2. Let for all $j \ge -1$, C_j be a subset of \mathbb{Z} . There is an event of probability $1 - 1/n^4$ on which: for all $j \ge -1$, all finite subset K_j of C_j ,

(26)
$$2^{-j/2} \sum_{k \in K_j} \left| \widehat{\beta}_{j,k} - \beta_{j,k} \right| \le c_j V_j(K_j, \mathcal{C}_j)$$

where

$$V_{j}(K_{j}, \mathcal{C}_{j}) = \sqrt{\frac{|K_{j}|\sigma_{j}^{2}(K_{j})\log_{+}(\sigma_{j}^{2}(\mathcal{C}_{j})/\sigma_{j}^{2}(K_{j}))}{n}} + \frac{|K_{j}|\log_{+}(n\sigma_{j}^{2}(\mathcal{C}_{j})/|K_{j}|)}{n} + \sqrt{\sigma_{j}^{2}(K_{j})\frac{\log((j+2)n)}{n}} + \frac{\log((j+2)n)}{n}.$$

Moreover, c_j is universal if j = -1 and only depends on ψ if $j \ge 0$.

Proof of Lemma 2. We only show the lemma when $j \ge 0$. The proof when j = -1 is similar (replace ψ by ϕ). We consider some $c_1 > 0$ only depending on ψ such that

$$\left\|\sum_{k\in\mathbb{Z}} \left|\psi(2^j\cdot -k)\right|\right\|_{\infty} \le 1/c_1.$$

We then consider $d \ge 1$ and define the at most countable collection

$$\mathcal{F}_j(d) = \left\{ c_1 \sum_{k \in K_j} \psi(2^j \cdot -k), \ K_j \subset \mathcal{C}_j, \ |K_j| \le d \right\}$$

`

of functions. These functions take values in [-1,1] and are piecewise constant on at most c_2d pieces where c_2 only depends on ψ . Therefore, $\mathcal{F}_i(d)$ is VC subgraph and its dimension is not larger than d, up to a multiplicative factor depending on ψ only (see [BB18] for instance).

Let now $K_j \subset \mathcal{C}_j$ such that $|K_j| \leq d$. Elementary computations entail

$$\sum_{k \in K_j} \left| \widehat{\beta}_{j,k} - \beta_{j,k} \right| \le 2 \sup_{K'_j \subset K_j} \left| \sum_{k \in K'_j} \left(\widehat{\beta}_{j,k} - \beta_{j,k} \right) \right|.$$

We introduce the map $\psi_{K'_j}(\cdot)$ defined for $x \in \mathbb{R}$ by

$$\psi_{K'_j}(x) = c_1 \sum_{k \in K'_j} \psi(2^j x - k).$$

The preceding inequality can then be rewritten as

$$2^{-j/2} \sum_{k \in K_j} \left| \widehat{\beta}_{j,k} - \beta_{j,k} \right| \le (2/c_1) \sup_{K'_j \subset K_j} \left| \frac{1}{n} \sum_{i=1}^n \left[\psi_{K'_j}(X_i) - \mathbb{E}(\psi_{K'_j}(X_i)) \right] \right|.$$

Note that

$$\bar{f}(x) = c_1 \sum_{k \in \mathcal{C}_j} \left| \psi(2^j x - k) \right|$$

is an envelope function of $\mathcal{F}_j(d)$. Moreover, Lemma 1 gives when $K'_j \subset K_j$,

$$\mathbb{E}[\psi_{K'_j}^2(X)] \le c_3 \sigma_j^2(K'_j) \le c_3 \sigma_j^2(K_j),$$

and $\mathbb{E}[\bar{f}^2(X)] \leq c_3 \sigma_j^2(\mathcal{C}_j)$. We apply the probabilistic result given by Proposition 6 in Appendix A.We then use that $x \mapsto x \log_+(a/x)$ is non-decreasing for all a > 0, and conclude by a union bound.

Lemma 3. Let for all $j \ge -1$, $r \in \mathbb{N}$, $\mathcal{C}_{j,r}$ be a finite subset of \mathbb{Z} . There is an event of probability $1-1/n^4$ on which: for all $j \ge -1, r \in \mathbb{N}$, all subset $K_{j,r}$ of $\mathcal{C}_{j,r}$,

(27)
$$2^{-j/2} \sum_{k \in K_{j,r}} \left| \widehat{\beta}_{j,k} - \beta_{j,k} \right| \le c_j W_{j,r}(K_{j,r}, \mathcal{C}_{j,r}),$$

where

$$W_{j,r}(K_{j,r}, \mathcal{C}_{j,r}) = \sqrt{\frac{|K_{j,r}|\sigma_j^2(K_{j,r})\log_+(|\mathcal{C}_{j,r}|/|K_{j,r}|)}{n}} + \frac{|K_{j,r}|\log_+(|\mathcal{C}_{j,r}|/|K_{j,r}|)}{n} + \sqrt{\sigma_j^2(K_{j,r})\frac{\log((j+2)(r+1)n)}{n}} + \frac{\log((j+2)(r+1)n)}{n}.$$

Moreover, c_j is universal if j = -1 and only depends on ψ if $j \ge 0$.

Sketch of the proof of Lemma 3. The proof is a slight variant to that of Lemma 2. The main difference is that we use (69) in Appendix A in place of (68). We restrict to the case $j \ge 0$, and define the collection $\mathcal{F}_{j,r}(d)$ for $d \in [1, |\mathcal{C}_{j,r}|]$ by

$$\mathcal{F}_{j,r}(d) = \left\{ c_1 \sum_{k \in K_{j,r}} \psi(2^j \cdot -k), \ K_{j,r} \subset \mathcal{C}_{j,r}, \ |K_{j,r}| \le d \right\}.$$

We have

$$|\mathcal{F}_{j,r}(d)| \le \sum_{i=0}^{d} \binom{|\mathcal{C}_{j,r}|}{i}.$$

Proposition 2.5 of [Mas07] entails $|\mathcal{F}_{j,r}(d)| \leq (e|\mathcal{C}_{j,r}|/d)^d$. The result follows from Proposition 6 and a union bound.

We omit the (easy) proof of the lemma below:

Lemma 4. Let $f_{j,\cdot} = (f_{j,k})_{k \in \mathbb{Z}}$ be defined by (16). Then, for all $j \ge -1$, $||f_{j,\cdot}||_1 \le c$, where c only depends on ψ if $j \ge 0$ and is universal if j = -1.

We now state:

Lemma 5. We define for all $j \ge -1$, $r \in \mathbb{N}$, $m \in \{0, 1, 2\}$, and all finite subset $K_{j,r}$ of \mathbb{Z} ,

$$\mathcal{E}_{j,r}(K_{j,r},m) = \sqrt{\frac{|K_{j,r}|\sigma_j^2(K_{j,r})\log_+(s_{j,r}(K_{j,r},m))}{n}} + \frac{|K_{j,r}|\log_+(s_{j,r}'(K_{j,r},m))}{n} + \sqrt{\sigma_j^2(K_{j,r})\frac{\log((j+2)(r+1)n)}{n}} + \frac{\log((j+2)(r+1)n)}{n},$$

where

$$\begin{aligned} s_{j,r}(K_{j,r},0) &= \mathbb{E}[|\widetilde{\mathbb{Z}}_{j}|]/|K_{j,r}| & and & s'_{j,r}(K_{j,r},0) &= \mathbb{E}[|\widetilde{\mathbb{Z}}_{j}|]/|K_{j,r}| \\ s_{j,r}(K_{j,r},1) &= 1/\sigma_{j}^{2}(K_{j,r}) & and & s'_{j,r}(K_{j,r},1) &= n/|K_{j,r}| \\ s_{j,r}(K_{j,r},2) &= |\breve{\mathbb{Z}}_{j,r}|/|K_{j,r}| & and & s'_{j,r}(K_{j,r},2) &= |\breve{\mathbb{Z}}_{j,r}|/|K_{j,r}|. \end{aligned}$$

Then, there is an event of probability $1-4/n^4$ on which: for all $j \ge -1$, $r \in \mathbb{N}$, all finite subset $K_{j,r}$ of \mathbb{Z} , and all $m \in \{0,1\}$,

(28)
$$2^{-j/2} \sum_{k \in K_{j,r}} \left| \widehat{\beta}_{j,k} - \beta_{j,k} \right| \le c_j \mathcal{E}_{j,r}(K_{j,r},m).$$

Moreover, if $K_{j,r} \subset \mathbb{Z}_{j,r}$ where $\mathbb{Z}_{j,r}$ is defined by (23), the left-hand side of (28) is also not larger than $c_j \mathcal{E}_{j,r}(K_{j,r}, 2)$. Here, c_j is universal if j = -1 and only depends on ψ if $j \ge 0$.

Proof of Lemma 5. We only show the lemma when $|K_{j,r}| \geq 1$. Note first that $\sigma_j^2(\mathbb{Z})$ is bounded from above by a numerical value when j = -1 and by a term only depending on ψ when $j \geq 0$ (use Lemma 4 and $\|\psi\|_{\infty} < \infty$). The proof that (28) holds true with m = 1 follows therefore from Lemma 2 (with $C_j = \mathbb{Z}$). The proof that it is also true with m = 2 is due to Lemma 3 (with $C_{j,r} = \check{\mathbb{Z}}_{j,r}$). We now suppose that m = 0. Let g be the map defined for $x \in [0,1]$ by $g(x) = 1 - (1-x)^n$. Note that $\hat{\sigma}_{j,k}^2 > 0$ if and only if $\psi_{j,k}(X_i) \neq 0$ for some i. Therefore,

(29)
$$\mathbb{E}[|\widehat{\mathbb{Z}}_j|] = \sum_{k \in \mathbb{Z}} g(f_{j,k}),$$

where $f_{j,k}$ is defined by (16) by $f_{j,k} = \int_{\text{supp }\psi_{j,k}} f$. Note that g is increasing and $x \mapsto g(x)/x$ is decreasing. Recall that $\overline{\mathbb{Z}}_j$ is defined in (15). We deduce,

$$\mathbb{E}[|\widehat{\mathbb{Z}}_{j}|] = \sum_{k \in \overline{\mathbb{Z}}_{j}} g(f_{j,k}) + \sum_{k \notin \overline{\mathbb{Z}}_{j}} \frac{g(f_{j,k})}{f_{j,k}} f_{j,k}$$
$$\geq |\overline{\mathbb{Z}}_{j}|g(1/n) + \sum_{k \notin \overline{\mathbb{Z}}_{j}} \frac{g(1/n)}{1/n} f_{j,k}.$$

Note that

$$\sigma_j^2(\overline{\mathbb{Z}}_j^c) \le \max\left\{1, \|\psi\|_{\infty}^2\right\} \sum_{k \notin \overline{\mathbb{Z}}_j} f_{j,k}$$

and hence

(30)
$$\mathbb{E}[|\widehat{\mathbb{Z}}_j|] \ge c_1 \left[|\overline{\mathbb{Z}}_j| + n\sigma_j^2(\overline{\mathbb{Z}}_j^c)\right].$$

where c_1 only depends on ψ . We deduce from Lemmas 2 and 3 that with probability $1 - 2/n^4$,

$$2^{-j/2} \sum_{k \in K_{j,r}} \left| \widehat{\beta}_{j,k} - \beta_{j,k} \right| \le c_2 \left[V_j(K_{j,r} \cap \overline{\mathbb{Z}}_j^c, \overline{\mathbb{Z}}_j^c) + W_{j,r}(K_{j,r} \cap \overline{\mathbb{Z}}_j, \overline{\mathbb{Z}}_j) \right].$$

Elementary computations give

$$V_j(K_{j,r} \cap \overline{\mathbb{Z}}_j^c, \overline{\mathbb{Z}}_j^c) \le c_3 \left[\frac{|K_{j,r}| \log_+(n\sigma_j^2(\overline{\mathbb{Z}}_j^c)/|K_{j,r}|)}{n} + \frac{\log((j+2)n)}{n} \right].$$

We conclude using (30).

Lemma 6. There is an event of probability $1 - 4/n^4$ on which: for all $j \ge -1$, $\varepsilon > 0$, and finite subset K_j of \mathbb{Z} ,

(31)
$$\widehat{\sigma}_j^2(K_j) \le (1+\varepsilon)\sigma_j^2(K_j) + c_j \left[\frac{|K_j|\log_+(\mathbb{E}[|\widehat{\mathbb{Z}}_j|]/|K_j|)}{n} + \frac{\log((j+2)n)}{n}\right]$$

(32)
$$\sigma_j^2(K_j) \le (1+\varepsilon)\widehat{\sigma}_j^2(K_j) + c_j \left[\frac{|K_j|\log_+(\mathbb{E}[|\widehat{\mathbb{Z}}_j|]/|K_j|)}{n} + \frac{\log((j+2)n)}{n}\right],$$

where c_j depends only on ε if j = -1 and only depends on ε, ψ if $j \ge 0$.

Sketch of the proof of Lemma 6. We may replace ψ in the previous proofs by ψ^2 . Hence,

$$\sum_{k \in K_j} \left| \widehat{\sigma}_{j,k}^2 - \sigma_{j,k}^2 \right| \le c_j' \mathcal{E}_{j,0}(K_j, 0),$$

where c'_j is universal if j = -1 and only depends on ψ if $j \ge 0$. We conclude using the elementary inequality $2\sqrt{xy} \le \alpha^{-1}x + \alpha y$ valid for all $\alpha > 0$.

Lemma 7. For all $j \ge -1$, $\mathbb{E}[|\widehat{\mathbb{Z}}_j|] \le cn$, where c only depends on ψ when $j \ge 0$ and is universal if j = -1.

Proof of Lemma 7. We suppose that $j \ge 0$ and note that $\hat{\sigma}_{j,k}^2 > c_1/n$ when $\hat{\sigma}_{j,k}^2 \ne 0$ as ψ is piecewise constant. Therefore,

$$\mathbb{E}\big[|\widehat{\mathbb{Z}}_j|\big] \le (n/c_1) \sum_{k \in \mathbb{Z}} \sigma_{j,k}^2.$$

Since ψ is bounded above, $\sigma_{j,k}^2 \leq c_2 f_{j,k}$. We then use Lemma 4.

Lemma 8. There exist a universal constant ς_{-1} , a term ς_0 only depending on ψ and an event of probability $1-4/n^4$ (the same as that of Lemma 6) on which: for all $j \ge -1$, $\ell_j \in \mathbb{N}$, $r \ge \ell_j + j + 1$ such that

$$2^r \leq \varsigma_j \frac{n}{\log((j+2)n)},$$

where $\varsigma_j = \varsigma_{-1} 1_{j=-1} + \varsigma_0 1_{j\geq 0}$, the embedding (22) holds true. We also have (21) on this event for all $j \geq -1$ and $k \in \mathbb{Z}$.

Proof of Lemma 8. The result follows from Lemma 6 with ε small enough and Lemma 7.

Lemma 9. We define for all $j \ge -1$, $r \in \mathbb{N}$, $\ell_j \in \mathbb{N}$, $m \in \{0, 1, 2\}$, and all finite subset K_{j,r,ℓ_j} of \mathbb{Z} ,

$$\widetilde{\mathcal{E}}_{j,r,\ell_j}(K_{j,r,\ell_j},m) = \sqrt{\frac{|K_{j,r,\ell_j}|\widehat{\sigma}_j^2(K_{j,r,\ell_j})\log_+(\widetilde{s}_{j,r,\ell_j}(K_{j,r,\ell_j}))}{n}} + \frac{|K_{j,r,\ell_j}|\log_+(\widetilde{s}_{j,r,\ell_j}(K_{j,r,\ell_j}))}{n} + \frac{\sqrt{\widehat{\sigma}_j^2(K_{j,r,\ell_j})\log((j+2)(r+1)n)}}{n} + \frac{\log((j+2)(r+1)n)}{n},$$

where

$$\begin{split} &\widetilde{s}_{j,r,\ell_{j}}(K_{j,r,\ell_{j}},0) = \mathbb{E}[|\widehat{\mathbb{Z}}_{j}|]/|K_{j,r,\ell_{j}}| & and & \widetilde{s}_{j,r,\ell_{j}}'(K_{j,r,\ell_{j}},0) = \mathbb{E}[|\widehat{\mathbb{Z}}_{j}|]/|K_{j,r,\ell_{j}}| \\ &\widetilde{s}_{j,r,\ell_{j}}(K_{j,r,\ell_{j}},1) = 1/\widehat{\sigma}_{j}^{2}(K_{j,r,\ell_{j}}) & and & \widetilde{s}_{j,r,\ell_{j}}'(K_{j,r,\ell_{j}},1) = n/|K_{j,r,\ell_{j}}| \\ &\widetilde{s}_{j,r,\ell_{j}}(K_{j,r,\ell_{j}},2) = |\widehat{\mathbb{Z}}'_{j,r}|/|K_{j,r,\ell_{j}}| & and & \widetilde{s}'_{j,r,\ell_{j}}(K_{j,r,\ell_{j}},2) = |\widehat{\mathbb{Z}}'_{j,r}|/|K_{j,r,\ell_{j}}|. \end{split}$$

Then, there is an event of probability $1 - 8/n^4$ on which: for all $j \ge -1$, $\ell_j \in \mathbb{N}$, $r \ge \ell_j + j$, finite subset K_{j,r,ℓ_j} of \mathbb{Z} , and $m \in \{0,1\}$,

$$2^{-j/2} \sum_{k \in K_{j,r,\ell_j}} \left| \widehat{\beta}_{j,k} - \beta_{j,k} \right| \le c_j \widetilde{\mathcal{E}}_{j,r,\ell_j}(K_{j,r,\ell_j},m).$$

Moreover, if $K_{j,r,\ell_j} \subset \widehat{\mathbb{Z}}_{j,r}(\ell_j)$ for some $r \geq \ell_j + j + 1$ and

$$2^r \leq \varsigma_j \frac{n}{\log((j+2)n)}$$

where ς_j is given by Lemma 8, the inequality holds true with m = 2. Furthermore, c_j is universal if j = -1 and only depends on ψ if $j \ge 0$.

Proof of Lemma 9. The case m = 0 is merely due to Lemma 5 and (32). The proof when m = 2 follows from Lemma 5, from the inclusions $\widehat{\mathbb{Z}}_{j,r}(\ell_j) \subset \mathbb{Z}_{j,r} \subset \widehat{\mathbb{Z}}'_{j,r}$ and from the double inequality

$$1/\sqrt{2} \le \frac{\sigma_j^2(K_{j,r,\ell_j})}{\widehat{\sigma}_j^2(K_{j,r,\ell_j})} \le \sqrt{2}$$

valid for all $K_{j,r,\ell_j} \subset \widehat{\mathbb{Z}}_{j,r}(\ell_j)$. The case m = 1 essentially follows from elementary computations: let σ^2 be the right-hand side of (32) (with $K_j = K_{j,r,\ell_j}$ and $\varepsilon = 1$). Then, using that $x \mapsto x \log_+(1/x)$ is non-decreasing,

$$\begin{aligned} \sigma_{j}^{2}(K_{j,r,\ell_{j}}) \log_{+}(1/\sigma_{j}^{2}(K_{j,r,\ell_{j}})) &\leq \sigma^{2} \log_{+}(1/\sigma^{2}) \\ &\leq c \left[\widehat{\sigma}_{j}^{2}(K_{j,r,\ell_{j}}) \log_{+}(1/\widehat{\sigma}_{j}^{2}(K_{j,r,\ell_{j}})) \right. \\ &\left. + (|K_{j,r,\ell_{j}}|/n) \log_{+}(\mathbb{E}[|\widehat{\mathbb{Z}}_{j}|]/|K_{j,r,\ell_{j}}|) \log_{+}(n/|K_{j,r,\ell_{j}}|) \right. \\ &\left. + (\log((j+2)n)/n) \log_{+}(n/|K_{j,r,\ell_{j}}|) \right]. \end{aligned}$$

Lemma 7 ends the proof.

Lemma 10. Lemma 9 holds true (up to an increase of c_j) with $\widetilde{s}'_{j,r,\ell_j}(K_{j,r,\ell_j},1)$ replaced by

$$\widetilde{s}'_{j,r,\ell_j}(K_{j,r,\ell_j},1) = 1/\widehat{\sigma}_j^2(K_{j,r,\ell_j}).$$

Proof of Lemma 10. We set

$$A_{j} = \sqrt{\frac{|K_{j,r,\ell_{j}}|\widehat{\sigma}_{j}^{2}(K_{j,r,\ell_{j}})\log_{+}(1/\widehat{\sigma}_{j}^{2}(K_{j,r,\ell_{j}}))}{n}}$$
$$B_{j} = \frac{|K_{j,r,\ell_{j}}|\log_{+}(1/\widehat{\sigma}_{j}^{2}(K_{j,r,\ell_{j}}))}{n}}{n}$$
$$B'_{j} = \frac{|K_{j,r,\ell_{j}}|\log_{+}(n/|K_{j,r,\ell_{j}}|)}{n}.$$

We observe that $x \mapsto x \log_+(1/x)$ is non-decreasing and $\sqrt{\log_+(x/\log_+(x))} \ge 0.8\sqrt{\log_+(x)}$. We deduce that if

$$\widehat{\sigma}_j^2(K_{j,r,\ell_j}) \ge \frac{|K_{j,r,\ell_j}|\log_+(n/|K_{j,r,\ell_j}|)}{n}$$

then $A_j \ge 0.8B'_j$. If now

$$\widehat{\sigma}_j^2(K_{j,r,\ell_j}) < \frac{|K_{j,r,\ell_j}|\log_+(n/|K_{j,r,\ell_j}|)}{n},$$

then $B_j \ge 0.8^2 B'_j$.

We therefore get some c > 0 such that

$$A_j + B'_j \le c(A_j + B_j),$$

which concludes the proof.

Lemma 11. For all $j \ge -1$, $\xi > 0$, and probability $1 - e^{-\xi}$,

$$\mathbb{E}\big[|\widehat{\mathbb{Z}}_j|\big] \le 2|\widehat{\mathbb{Z}}_j| + 2(2L_{\psi}+1)\xi.$$

Proof of Lemma 11. This result derives from a Poissonian inequality for self-bounding functionals and more precisely from equation (7) of [BLM00]. We set for $k \in \mathbb{Z}$,

$$I_j(k) = \{x \in \mathbb{R}, \psi_{j,k}(x) \neq 0\}$$
$$\subset \left[2^{-j}(k - L_{\psi}), 2^{-j}(k + L_{\psi})\right],$$

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and remark

$$|\widehat{\mathbb{Z}}_j| = \sum_{k \in \mathbb{Z}} \mathbb{1}_{\exists i \in \{1,\dots,n\}, X_i \in I_j(k)}.$$

We introduce for $r \in \{1, ..., n\}$ the random variable

$$|\widehat{\mathbb{Z}}_{j}(r)| = \sum_{k \in \mathbb{Z}} 1_{\exists i \in \{1, \dots, n\} \setminus \{r\}, X_{i} \in I_{j}(k)}.$$

We have $|\widehat{\mathbb{Z}}_j(r)| \leq |\widehat{\mathbb{Z}}_j|$. Moreover,

$$\begin{aligned} \widehat{\mathbb{Z}}_{j}|-|\widehat{\mathbb{Z}}_{j}(r)| &= \sum_{k \in \mathbb{Z}} \mathbb{1}_{X_{r} \in I_{j}(k)} \mathbb{1}_{\forall i \in \{1,...,n\} \setminus \{r\}, \ X_{i} \notin I_{j}(k)} \\ &\leq \sum_{k \in \mathbb{Z}} \mathbb{1}_{X_{r} \in I_{j}(k)} \\ &\leq 2L_{\psi} + 1. \end{aligned}$$

Besides,

$$\sum_{r=1}^{n} \left(|\widehat{\mathbb{Z}}_{j}| - |\widehat{\mathbb{Z}}_{j}(r)| \right) = \sum_{k \in \mathbb{Z}} \sum_{r=1}^{n} 1_{X_{r} \in I_{j}(k)} \mathbb{1}_{\forall i \in \{1, \dots, n\} \setminus \{r\}, \ X_{i} \notin I_{j}(k)}$$
$$\leq \sum_{k \in \mathbb{Z}} \mathbb{1}_{\exists i \in \{1, \dots, n\}, \ X_{i} \in I_{j}(k)}$$
$$\leq |\widehat{\mathbb{Z}}_{j}|.$$

Now, equation (7) of [BLM00] gives for all $\xi > 0$, and probability $1 - e^{-\xi}$,

$$\mathbb{E}\big[|\widehat{\mathbb{Z}}_j|\big] \le |\widehat{\mathbb{Z}}_j| + \sqrt{2(2L_{\psi}+1)\mathbb{E}\big[|\widehat{\mathbb{Z}}_j|\big]\xi}$$

We conclude by using the elementary inequality $\sqrt{ab} \leq a/2 + b/2$.

Lemma 12. For all $j \ge -1$, $\xi > 0$, and probability $1 - e^{-\xi}$,

$$|\widehat{\mathbb{Z}}_j| \le 2\mathbb{E}\big[|\widehat{\mathbb{Z}}_j|\big] + (7/6)(2L_{\psi}+1)\xi.$$

Proof of Lemma 12. The proof is similar to that of Lemma 12. We merely use (6) of [BLM00] to get for all $\xi > 0$, and probability $1 - e^{-\xi}$,

$$|\widehat{\mathbb{Z}}_j| \leq \mathbb{E}\big[|\widehat{\mathbb{Z}}_j|\big] + \sqrt{2(2L_{\psi}+1)\mathbb{E}\big[|\widehat{\mathbb{Z}}_j|\big]\xi} + 2\frac{2L_{\psi}+1}{3}\xi.$$

We conclude as in the preceding proof.

Lemma 13. With probability $1 - 1/n^4$, we have for all $j \ge -1$,

$$\mathbb{E}\left[|\widehat{\mathbb{Z}}_j|\right] \le 2|\widehat{\mathbb{Z}}_j| + 4(2L_{\psi}+1)\left(\log(j+2) + 2\log n + 1\right),$$

and

$$|\widehat{\mathbb{Z}}_j| \le 2\mathbb{E}[|\widehat{\mathbb{Z}}_j|] + (7/3)(2L_{\psi} + 1)(\log(j+2) + 2\log n + 1).$$

Proof of Lemma 13. The proof follows from Lemmas 11, 12 and a union bound.

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Lemma 14. There exist a universal constant $\bar{\varrho}_{-1}$, a term $\bar{\varrho}_0$ depending only on ψ , and an event of probability $1-9/n^4$ on which: for all $j \ge -1$, $\ell_j \in \mathbb{N}$, $r \ge \ell_j + j$, and finite subset K_{j,r,ℓ_j} of $\widehat{\mathbb{Z}}_{j,r}(\ell_j)$,

$$2^{-j/2} \sum_{k \in K_{j,r,\ell_j}} \left| \widehat{\beta}_{j,k} - \beta_{j,k} \right| \le (\overline{\varrho}_j/2) \widehat{\mathcal{E}}_{j,r,\ell_j}(|K_{j,r,\ell_j}|),$$

where $\bar{\varrho}_j = \bar{\varrho}_0$ if $j \ge 1$. Moreover, (20), (21), and (22) hold true on this event.

Proof of Lemma 14. The lemma is a direct result of Lemmas 8, 9, 10 and 13 when $r \ge \ell_j + j + 1$ (the event of Lemma 8 contains the one of Lemma 9). When $r = \ell_j + j$, we use

$$\widehat{\sigma}_j^2(K_{j,r,\ell_j}) \le \bar{\kappa}_j(2L_\psi + 1),$$

with $\bar{\kappa}_j = 1_{j=-1} + \|\psi\|_{\infty}^2 1_{j\geq 0}$. Moreover,

$$\widehat{\sigma}_j^2(K_{j,r,\ell_j}) \ge 2^{-r-1} |K_{j,r,\ell_j}|.$$

We conclude by using Lemma 9 and the fact that $x \mapsto x \log_+(1/x)$ is non-decreasing.

Lemma 15. There exist terms $\varsigma_0, \bar{\varrho}_0$ only depending on ψ and universal constants $\varsigma_{-1}, \bar{\varrho}_{-1}$ such that if $\rho_{-1} \geq \bar{\varrho}_{-1}$ and $\rho_0 \geq \bar{\varrho}_0$, the estimator \hat{f} defined by (6) satisfies with probability $1 - 9/n^4$: for all $(\ell_j)_{j\geq -1} \in \mathbb{N}^{\{-1\}\cup\mathbb{N}}$,

$$\begin{split} \left\| f - \hat{f} \right\|_1 &\leq c \widehat{T} \\ &+ c \sum_{j=-1}^{\widehat{J}} \sum_{r=\ell_j+j}^{\infty} \inf_{K_{j,r,\ell_j} \subset \widehat{\mathbb{Z}}_{j,r}(\ell_j) \cap \overline{\mathbb{Z}}_j} \left\{ 2^{-j/2} \sum_{k \in (\widehat{\mathbb{Z}}_{j,r}(\ell_j) \cap \overline{\mathbb{Z}}_j) \setminus K_{j,r,\ell_j}} |\beta_{j,k}| + \rho_j \widehat{\mathcal{E}}_{j,r,\ell_j}(|K_{j,r,\ell_j}|) \right\}, \end{split}$$

where

(33)
$$\widehat{T} = \sum_{j=-1}^{\widehat{J}} 2^{-j/2} \sum_{k \notin \widehat{\mathbb{Z}}_j} |\beta_{j,k}| + \sum_{j=-1}^{\widehat{J}} 2^{-j/2} \sum_{k \notin \overline{\mathbb{Z}}_j} |\beta_{j,k}| + \sum_{j=\widehat{J}+1}^{\infty} 2^{-j/2} \sum_{k \in \mathbb{Z}} |\beta_{j,k}|,$$

where $\overline{\mathbb{Z}}_j$ is given by (15) and where c only depends on $\overline{\phi}, \overline{\psi}$. Moreover, (20), (21), and (22) hold true on this event.

Proof of Lemma 15. We observe that for all $j \geq -1$ and $\ell_j \in \mathbb{N}$,

(34)
$$\widehat{\mathbb{Z}}_{j} = \bigcup_{r=\ell_{j}+j}^{\infty} \widehat{\mathbb{Z}}_{j,r}(\ell_{j})$$

In particular,

$$\mathbb{Z} = \bigcup_{r=\hat{\ell}_j+j}^{\infty} \ \widehat{\mathbb{Z}}_{j,r}(\hat{\ell}_j) \bigcup \widehat{\mathbb{Z}}_j^c.$$

We deduce from (2) that $\left\|f - \hat{f}\right\|_1 \le c\widetilde{T} + c\widehat{A}$ where \widehat{z}

$$\widetilde{T} = \sum_{j=-1}^{\widehat{J}} 2^{-j/2} \sum_{k \notin \widehat{\mathbb{Z}}_j} |\beta_{j,k}| + \sum_{j=\widehat{J}+1}^{\infty} 2^{-j/2} \sum_{k \in \mathbb{Z}} |\beta_{j,k}|$$
$$\widehat{A} = \sum_{j=-1}^{\widehat{J}} 2^{-j/2} \sum_{r=\widehat{\ell}_j+j}^{\infty} \left\{ \sum_{k \in \widehat{\mathbb{Z}}_{j,r}(\widehat{\ell}_j) \setminus \widehat{K}_{j,r,\widehat{\ell}_j}} |\beta_{j,k}| + \sum_{k \in \widehat{K}_{j,r,\widehat{\ell}_j}} \left| \widehat{\beta}_{j,k} - \beta_{j,k} \right| \right\}.$$

Note that

$$\widehat{A} \le \sum_{j=-1}^{\widehat{J}} 2^{-j/2} \left\{ \sum_{r=\hat{\ell}_j+j}^{\infty} \sum_{k \in \widehat{\mathbb{Z}}_{j,r}(\hat{\ell}_j)} |\beta_{j,k}| + \sum_{r=\hat{\ell}_j+j}^{\infty} \left\{ -\sum_{k \in \widehat{K}_{j,r,\hat{\ell}_j}} |\beta_{j,k}| + \sum_{k \in \widehat{K}_{j,r,\hat{\ell}_j}} \left| \widehat{\beta}_{j,k} - \beta_{j,k} \right| \right\} \right\}.$$

The triangle inequality and Lemma 14 entail: on an event of probability $1 - 9/n^4$,

$$- 2^{-j/2} \sum_{k \in \widehat{K}_{j,r,\hat{\ell}_{j}}} |\beta_{j,k}| + 2^{-j/2} \sum_{k \in \widehat{K}_{j,r,\hat{\ell}_{j}}} \left| \widehat{\beta}_{j,k} - \beta_{j,k} \right|$$

$$\leq -2^{-j/2} \sum_{k \in \widehat{K}_{j,r,\hat{\ell}_{j}}} |\widehat{\beta}_{j,k}| + 2^{1-j/2} \sum_{k \in \widehat{K}_{j,r,\hat{\ell}_{j}}} \left| \widehat{\beta}_{j,k} - \beta_{j,k} \right|$$

$$\leq -2^{-j/2} \sum_{k \in \widehat{K}_{j,r,\hat{\ell}_{j}}} |\widehat{\beta}_{j,k}| + \bar{\varrho}_{j} \widehat{\mathcal{E}}_{j,r,\hat{\ell}_{j}} (|\widehat{K}_{j,r,\hat{\ell}_{j}}|).$$

By gathering all these results, and by using $\rho_j \geq \bar{\varrho}_j$,

$$\widehat{A} \leq \sum_{j=-1}^{\widehat{J}} 2^{-j/2} \sum_{k \in \widehat{\mathbb{Z}}_j} |\beta_{j,k}| + \sum_{j=-1}^{\widehat{J}} \sum_{r=\hat{\ell}_j+j}^{\infty} \gamma_{j,r,\hat{\ell}_j} (\widehat{K}_{j,r,\hat{\ell}_j}).$$

We use (9), (8), triangle inequality and Lemma 14. This leads to the two following inequalities valid for all $\ell_j \in \mathbb{N}$:

$$\begin{split} \widehat{A} &\leq \sum_{j=-1}^{\widehat{J}} 2^{-j/2} \sum_{k \in \widehat{\mathbb{Z}}_{j}} |\beta_{j,k}| + \sum_{j=-1}^{\widehat{J}} \sum_{r=\ell_{j}+j}^{\infty} \inf_{\substack{K_{j,r,\ell_{j}} \subset \widehat{\mathbb{Z}}_{j,r}(\ell_{j})}} \left\{ -2^{-j/2} \sum_{k \in K_{j,r,\ell_{j}}} |\widehat{\beta}_{j,k}| + \rho_{j} \widehat{\mathcal{E}}_{j,r,\ell_{j}}(|K_{j,r,\ell_{j}}|) \right\} \\ &\leq \sum_{j=-1}^{\widehat{J}} 2^{-j/2} \sum_{k \in \widehat{\mathbb{Z}}_{j}} |\beta_{j,k}| + \sum_{j=-1}^{\widehat{J}} \sum_{r=\ell_{j}+j}^{\infty} \inf_{\substack{K_{j,r,\ell_{j}} \subset \widehat{\mathbb{Z}}_{j,r}(\ell_{j})}} \left\{ -2^{-j/2} \sum_{k \in K_{j,r,\ell_{j}}} |\beta_{j,k}| + (3/2)\rho_{j} \widehat{\mathcal{E}}_{j,r,\ell_{j}}(|K_{j,r,\ell_{j}}|) \right\} \\ &\text{We use (34) and the triangle inequality to conclude.} \qquad \Box$$

We use (34) and the triangle inequality to conclude.

Lemma 16. Let \hat{T} be defined by (33). Then,

$$\mathbb{E}[\widehat{T}] \le 3T_1 + 2T_2,$$

where T_1 and T_2 are defined by (18) and (19).

Proof of Lemma 16. There is at most one observation X_i such that $\psi_{j,k}(X_i) \neq 0$ when $j \geq \widehat{J} + 1$. Moreover, no observation X_i satisfies $\psi_{j,k}(X_i) \neq 0$ when $k \notin \widehat{\mathbb{Z}}_j$, no matter j. We deduce,

$$\mathbb{E}[\widehat{T}] \le T_1 + \sum_{j=-1}^{\infty} 2^{-j/2} \sum_{k \in \mathbb{Z}} |\beta_{j,k}| \left((1 - f_{j,k})^n + n f_{j,k} (1 - f_{j,k})^{n-1} \right),$$

where $f_{j,k}$ is given by (16). We conclude by noticing that $(1 - f_{j,k})^n + nf_{j,k}(1 - f_{j,k})^{n-1} \le 2$ if $f_{j,k} \le 1/n$ and $(1 - f_{j,k})^n \le nf_{j,k}(1 - f_{j,k})^{n-1}$ if $f_{j,k} > 1/n$.

Lemma 17. Let $\xi > 0$. The following assertion holds true with probability $1 - \xi/n$: for all q > 1 and $f \in \mathbb{L}^q$,

$$2^{\widehat{J}} \leq \max\left\{1, 8L_{\psi}\left(n^{2} \|f\|_{q}/\xi\right)^{\frac{q}{q-1}}\right\}.$$

In particular, for all $k \ge 1$, $\mathbb{E}[\widehat{J}^k] \le C \log^k n$ where C depends on k, q, $||f||_q$ and L_{ψ} only.

Proof of Lemma 17. The proof of this lemma is deferred to Appendix D.

Proof of Theorem 5. There exists c_1 only depending on $\overline{\phi}$, $\overline{\psi}$ such that

$$\|\hat{f}\|_{1} \leq \frac{c_{1}}{n} \sum_{i=1}^{n} \left[\sum_{k \in \mathbb{Z}} \left| \phi(2^{j}X_{i} - k) \right| + \sum_{j=0}^{\widehat{J}} \sum_{k \in \mathbb{Z}} \left| \psi(2^{j}X_{i} - k) \right| \right]$$

$$\leq c_{2}(\widehat{J} + 2).$$

We deduce from Lemma 17 an event \mathcal{A}_1 of probability $1 - 1/n^4$ on which $\widehat{J} \leq c_0 \log n$. This lemma also gives $\mathbb{E}[\widehat{J}^2] \leq c_3 \log^2 n$ and hence $\mathbb{E}[\|\widehat{f}\|_1^2] \leq c_4 \log^2 n$.

Let \mathcal{A}_2 be the event of probability $1 - 9/n^4$ that appears in Lemma 15. We set $\mathcal{A} = \mathcal{A}_1 \cap \mathcal{A}_2$ and get

$$\mathbb{E}\left[\|f-\hat{f}\|_{1}\right] \leq \mathbb{E}\left[\|f-\hat{f}\|_{1}\mathbf{1}_{\mathcal{A}}\right] + \mathbb{E}\left[\|f-\hat{f}\|_{1}\mathbf{1}_{\mathcal{A}^{c}}\right].$$

The first term can be bounded from above by using Lemmas 15 and 16. As to the second term, we use the triangle and Cauchy-Schwarz inequalities to get

$$\mathbb{E}\left[\|f - \hat{f}\|_{1} \mathbb{1}_{\mathcal{A}^{c}}\right] \leq \mathbb{P}\left(\mathcal{A}^{c}\right) + \mathbb{E}\left[\|\hat{f}\|_{1}^{2}\right]^{1/2} \mathbb{P}\left(\mathcal{A}^{c}\right)^{1/2}$$
$$\leq c_{5} \frac{\log n}{n^{2}}.$$

4.2. Proof of Theorem 2: intermediate lemmas. We first recall the following result. We refer to [CVNRF15] for its proof (see their Propositions 3.2 and 4.5).

Lemma 18. Let Λ be an at most countable set, and $x = (x_{\lambda})_{\lambda \in \Lambda} \in \mathbb{R}^{\Lambda}_+$. For all 0 ,

(35)
$$\|x\|_{q}^{q} \leq \frac{q}{q-p} \|x\|_{\infty}^{q-p} \|x\|_{p,\infty}^{p}$$

Moreover, for all p > 1, and finite subset Γ of Λ ,

(36)
$$\sum_{\lambda \in \Gamma} x_{\lambda} \leq \frac{p}{p-1} \|x\|_{p,\infty} \|\Gamma\|^{1-1/p}.$$

The lemma below is elementary.

Lemma 19. Suppose that $f \in WT_{\theta}(M)$ for some $\theta \in (0,1)$ and let $f_{j,\cdot} = (f_{j,k})_{k \in \mathbb{Z}}$ be defined by (16). Then, for all $j \geq -1$,

$$\|f_{j,\cdot}\|_{\theta,\infty}^{\theta} \le cM2^{j(1-\theta)},$$

where c only depends on ψ and θ if $j \ge 0$ and only depends on θ if j = -1.

We now show:

Lemma 20. Let $p \in (0, \infty]$, $\alpha \in ((1/p-1)_+, \tau)$, $\theta \in [0,1) \cap [0,p]$, R > 0, $M \ge 1$. Consider $v_n > 0$ and define

$$T_1(v_n) = \sum_{j=-1}^{\infty} 2^{-j/2} \sum_{\substack{k \in \mathbb{Z} \\ f_{j,k} \le v_n}} |\beta_{j,k}|.$$

Then, for all $f \in \mathscr{F}_{\alpha,p,\theta}(R,M)$,

(37)
$$T_1(v_n) \le c \left[R^{(1-\theta)/(1+\alpha-\theta/p)} M^{(\alpha+1-1/p)/(\alpha+1-\theta/p)} v_n^{\alpha(1-\theta)/(\alpha+1-\theta/p)} + M v_n^{1-\theta} \right],$$

where c only depends on p, α, θ, ψ .

Proof of Lemma 20. We define for $r \ge 0, j \ge -1$, $\overline{\mathbb{Z}}_{j,r} = \left\{ k \in \mathbb{Z}, 2^{-r-1} < f_{j,k} \le 2^{-r} \right\}.$

Since $f \in \mathcal{WT}_{\theta}(M)$, Lemma 19 implies when $\theta \in (0, 1)$,

(38)
$$|\overline{\mathbb{Z}}_{j,r}| \leq 2^{(r+1)\theta} \|f_{j,\cdot}\|_{\theta,\infty}^{\theta} \leq c_1 M 2^{r\theta} 2^{j(1-\theta)},$$

where c_1 only depends on ψ . We can check that this result remains true when $\theta = 0$. We also remark that $2^{-j/2}|\beta_{j,k}| \leq c_2 f_{j,k}$ as ψ is bounded and hence

$$T_1(v_n) \le A + T_1'(v_n),$$

where

$$A = c_2 \sum_{2^r \ge 1/v_n} 2^{-r} |\overline{\mathbb{Z}}_{-1,r}|$$
$$T_1'(v_n) = \sum_{j=0}^{\infty} \sum_{2^r \ge 1/v_n} \sum_{k \in \overline{\mathbb{Z}}_{j,r}} \min\left\{ c_2 2^{-r}, 2^{-j/2} |\beta_{j,k}| \right\}.$$

Note that

$$A \le c_1 c_2 M 2^{-(1-\theta)} \sum_{2^r \ge 1/v_n} 2^{-r(1-\theta)} \le c_3 M v_n^{1-\theta}.$$

We now focus on $T'_1(v_n)$.

We first suppose that p > 1. By using (36),

$$2^{-j/2} \sum_{k \in \overline{\mathbb{Z}}_{j,r}} |\beta_{j,k}| \le c_4 2^{-j/2} \|\beta_{j,\cdot}\|_{p,\infty} |\overline{\mathbb{Z}}_{j,r}|^{1-1/p},$$

and $f \in \mathcal{WB}^{\alpha}_{p,\infty}(R)$, we get

$$T_{1}'(v_{n}) \leq c_{5} \sum_{j=0}^{\infty} \sum_{2^{r} \geq 1/v_{n}} \min\left\{2^{-r} |\overline{\mathbb{Z}}_{j,r}|, R2^{-j(\alpha+1-1/p)} |\overline{\mathbb{Z}}_{j,r}|^{1-1/p}\right\}$$
$$\leq c_{6} \sum_{j=0}^{\infty} \sum_{2^{r} \geq 1/v_{n}} \min\left\{M2^{-(r-j)(1-\theta)}, RM^{1-1/p}2^{r(1-1/p)\theta}2^{-j[(1-1/p)\theta+\alpha]}\right\}$$
$$\leq c_{7} R^{(1-\theta)/(1+\alpha-\theta/p)} M^{(\alpha+1-1/p)/(\alpha+1-\theta/p)} \sum_{2^{r} \geq 1/v_{n}} 2^{-r\alpha(1-\theta)/(\alpha+1-\theta/p)}$$
$$\leq c_{8} R^{(1-\theta)/(1+\alpha-\theta/p)} M^{(\alpha+1-1/p)/(\alpha+1-\theta/p)} v_{n}^{\alpha(1-\theta)/(\alpha+1-\theta/p)}.$$

We now suppose $p \leq 1$ and consider $j_0 \in \mathbb{N}$. We have $T'_1(\upsilon_n) \leq T'_{1,1}(\upsilon_n) + T'_{1,2}(\upsilon_n)$ where

$$T'_{1,1}(\upsilon_n) = \sum_{j=0}^{j_0-1} 2^{-j/2} \sum_{\substack{k \in \mathbb{Z} \\ f_{j,k} \le \upsilon_n}} |\beta_{j,k}|$$
$$T'_{1,2}(\upsilon_n) = \sum_{j=j_0}^{\infty} 2^{-j/2} \sum_{\substack{k \in \mathbb{Z} \\ f_{j,k} \le \upsilon_n}} |\beta_{j,k}|.$$

By using $2^{-j/2}|\beta_{j,k}| \leq c_2 f_{j,k}$, we get when $\theta \neq 0$,

$$T'_{1,1}(v_n) \leq c_9 \sum_{j=0}^{j_0-1} \sum_{\substack{k \in \mathbb{Z} \\ f_{j,k} \leq v_n}} f_{j,k}$$
$$\leq c_{10} v_n^{1-\theta} \sum_{j=0}^{j_0-1} \|f_{j,\cdot}\|_{\theta,\infty}^{\theta} \quad \text{thanks to (35)}$$
$$\leq c_{11} v_n^{1-\theta} M 2^{j_0(1-\theta)} \quad \text{thanks to Lemma 19.}$$

This last inequality remains true when $\theta = 0$. Moreover, by using (35) when $p \neq 1$,

$$T'_{1,2} \le c_{12} v_n^{1-p} \sum_{j=j_0}^{\infty} 2^{-jp/2} \|\beta_{j,\cdot}\|_{p,\infty}^p$$

$$\le c_{13} v_n^{1-p} \sum_{j=j_0}^{\infty} 2^{-jp/2} R^p 2^{-jp(\alpha+1/2-1/p)} \quad \text{as } f \in \mathcal{WB}_{p,\infty}^{\alpha}(R)$$

Note that this inequality also holds true when p = 1 and $f \in \mathcal{B}^{\alpha}_{1,\infty}(R)$. Therefore, in both cases,

$$T'_{1,2} \le c_{14} v_n^{1-p} R^p 2^{-j_0 p(\alpha+1-1/p)}.$$

We conclude by choosing j_0 appropriately.

Lemma 21. Let $p \in (0,\infty]$, $\alpha \in ((1/p-1)_+, \tau)$, $\theta \in [0,1) \cap [0,p]$, R > 0, $M \ge 1$. Then, for all $f \in \mathscr{F}_{\alpha,p,\theta}(R,M)$,

(39)
$$T_{1} + T_{2} \leq c \left[R^{(1-\theta)/(1+\alpha-\theta/p)} M^{(\alpha+1-1/p)/(\alpha+1-\theta/p)} n^{-\alpha(1-\theta)/(\alpha+1-\theta/p)} + M n^{-(1-\theta)} \right],$$

where T_1 and T_2 are defined by (18) and (19), and where c only depends on p, α, θ, ψ .

Proof of Lemma 21. We focus on T_2 as a bound on T_1 may be obtained via the preceding lemma. We first consider the case $p \ge 1$. Since $f \in \mathcal{WT}_{\theta}(M)$, we have when $\theta \ne 0$.

(40)
$$|\overline{\mathbb{Z}}_j| \le n^{\theta} \|f_{j,\cdot}\|_{\theta,\infty}^{\theta} \le c_1 M n^{\theta} 2^{j(1-\theta)},$$

where c_1 only depends on ψ (Lemma 19). This inequality remains true when $\theta = 0$.

We consider $j_0 \in \mathbb{N}$ and decompose T_2 as $T_2 = T_{2,1} + T_{2,2}$ where

$$T_{2,1} = n \sum_{j=-1}^{j_0-1} 2^{-j/2} \sum_{k \in \overline{\mathbb{Z}}_j} |\beta_{j,k}| f_{j,k} (1 - f_{j,k})^{n-1},$$

$$T_{2,2} = n \sum_{j=j_0}^{\infty} 2^{-j/2} \sum_{k \in \overline{\mathbb{Z}}_j} |\beta_{j,k}| f_{j,k} (1 - f_{j,k})^{n-1}.$$

We use $2^{-j/2}|\beta_{j,k}| \leq c_2 f_{j,k}$ to get

$$T_{2,1} \le c_2 n \sum_{j=-1}^{j_0-1} \sum_{k \in \overline{\mathbb{Z}}_j} f_{j,k}^2 (1-f_{j,k})^{n-1}.$$

Since $x^2(1-x)^{n-1} \le c_3/n^2$ for all $x \in [0,1]$,

(41)

$$T_{2,1} \leq \frac{c_4}{n} \sum_{j=-1}^{j_0-1} |\overline{\mathbb{Z}}_j|$$

$$\leq c_5 M n^{-(1-\theta)} 2^{j_0(1-\theta)} \quad \text{thanks to (40)}.$$

As to $T_{2,2}$, we deduce from (36), $f \in \mathcal{WB}^{\alpha}_{p,\infty}(R)$, and (40) that if p > 1,

$$2^{-j/2} \sum_{k \in \overline{\mathbb{Z}}_j} |\beta_{j,k}| \le c_6 R M^{1-1/p} 2^{-j(\alpha+\theta(1-1/p))} n^{(1-1/p)\theta},$$

and hence, using that $x(1-x)^{n-1} \leq c_7/n$,

$$T_{2,2} \le c_8 R M^{1-1/p} 2^{-j_0(\alpha+\theta(1-1/p))} n^{(1-1/p)\theta}$$

We finally choose j_0 in a suitable way to conclude the proof when p > 1. Note that the above reasoning also works with p = 1 if we replace the Lorentz norm $\|\cdot\|_{1,\infty}$ by the \mathbb{L}^1 norm.

We now turn to the case p < 1. We write $T_2 = T_{2,1} + T_{2,2}$ where

$$T_{2,1} = n \sum_{j=-1}^{j_1} 2^{-j/2} \sum_{k \in \overline{\mathbb{Z}}_j} |\beta_{j,k}| f_{j,k} (1 - f_{j,k})^{n-1}$$
$$T_{2,2} = n \sum_{j=j_1}^{\infty} 2^{-j/2} \sum_{k \in \overline{\mathbb{Z}}_j} |\beta_{j,k}| f_{j,k} (1 - f_{j,k})^{n-1}.$$

Note that (41) does not use p > 1 and therefore also holds true when p < 1 (with j_0 replaced by j_1). Besides, as $2^{-j/2}|\beta_{j,k}| \leq c_2 f_{j,k}$ and $x^2(1-x)^{n-1} \leq c_3/n^2$,

$$n2^{-j/2}|\beta_{j,k}|f_{j,k}(1-f_{j,k})^{n-1} \le c_2 n f_{j,k}^2 (1-f_{j,k})^{n-1} \le c_9/n.$$

We deduce from (35),

$$T_{2,2} \le c_{10} n^{-(1-p)} \sum_{j=j_1}^{\infty} \left\| n 2^{-j/2} \beta_{j,\cdot} f_{j,\cdot} (1-f_{j,\cdot})^{n-1} \right\|_{p,\infty}^p$$

Yet, $x(1-x)^{n-1} \le c_7/n$ for all $x \in [0,1]$ and hence,

$$T_{2,2} \le c_{11} n^{-(1-p)} \sum_{j=j_1}^{\infty} 2^{-jp/2} \|\beta_{j,\cdot}\|_{p,\infty}^p$$
$$\le c_{12} n^{-(1-p)} R^p 2^{-j_1 p(\alpha+1-1/p)}.$$

It then remains to choose j_1 to conclude.

Lemma 22. For all $j \ge -1$, $\ell_j \in \mathbb{N}$, $r \ge \ell_j + j$, $\theta \in [0,1)$, $M \ge 1$, $f \in \mathcal{WT}_{\theta}(M)$,

(42)
$$\mathbb{E}\left[|\widehat{\mathbb{Z}}_{j,r}(\ell_j) \cap \overline{\mathbb{Z}}_j|\right] \le cM2^{r\theta}2^{j(1-\theta)},$$

where c only depends on ψ and θ .

Proof of Lemma 22. The proof is straightforward when $\theta = 0$ and we assume therefore that $\theta > 0$. We have,

$$\mathbb{E}\left[\left|\widehat{\mathbb{Z}}_{j,r}(\ell_j) \cap \overline{\mathbb{Z}}_j\right|\right] \le \sum_{\substack{k \in \mathbb{Z} \\ f_{j,k} \ge 1/n}} \mathbb{P}\left[\widehat{\sigma}_{j,k}^2 > 2^{-r-1}\right].$$

Set $t = 2^{-r-1}$ and define

$$K_t = \left\{ k \in \mathbb{Z}, \, \sigma_{j,k}^2 \geq t/2 \right\}.$$

We derive from $\sigma_{j,k}^2 \leq \max\{1, \|\psi\|_{\infty}^2\} f_{j,k}$, Lemma 19, and $f \in \mathcal{WT}_{\theta}(M)$, that $|K_t| \leq c_1 M 2^{j(1-\theta)} t^{-\theta}$. Moreover,

$$\begin{split} \mathbb{E}\left[|\widehat{\mathbb{Z}}_{j,r}(\ell_j) \cap \overline{\mathbb{Z}}_j|\right] &\leq |K_t| + \sum_{\substack{k \notin K_t \\ f_{j,k} \geq 1/n}} \mathbb{P}\left[\widehat{\sigma}_{j,k}^2 \geq t\right] \\ &\leq c_1 M 2^{j(1-\theta)} t^{-\theta} + \sum_{\substack{k \notin K_t \\ f_{j,k} \geq 1/n}} \mathbb{P}\left[\widehat{\sigma}_{j,k}^2 \geq \sigma_{j,k}^2 + t/2\right] \end{split}$$

We use Bennett's inequality (and more precisely equation (2.16) of [Mas07]) to get

$$\mathbb{P}\left[\widehat{\sigma}_{j,k}^2 \ge \sigma_{j,k}^2 + t/2\right] \le \exp\left[-c_2 n t^2 / (\sigma_{j,k}^2 + t)\right],$$

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where c_2 only depends on $\|\psi\|_{\infty}$. Therefore,

$$\mathbb{E}\left[\left|\widehat{\mathbb{Z}}_{j,r}(\ell_j) \cap \overline{\mathbb{Z}}_j\right|\right] \le c_1 2^{j(1-\theta)} M t^{-\theta} + \sum_{\substack{k \notin K_t \\ f_{j,k} \ge 1/n}} e^{-c_3 n t}.$$

Note that the number of k such that $f_{j,k} \ge 1/n$ is bounded from above by $c_4 M n^{\theta} 2^{j(1-\theta)}$. Therefore,

$$\mathbb{E}\left[|\widehat{\mathbb{Z}}_{j,r}(\ell_j) \cap \overline{\mathbb{Z}}_j|\right] \le c_5 2^{j(1-\theta)} M t^{-\theta} \left[1 + (nt)^{\theta} e^{-c_3 nt}\right]$$

We conclude by remarking that the map $x \mapsto x^{\theta} e^{-c_3 x}$ is bounded on \mathbb{R} .

Lemma 23. For all $j \ge -1$, $\theta \in [0,1)$, $M \ge 1$, $f \in \mathcal{WT}_{\theta}(M)$,

(43)
$$\mathbb{E}\left[|\widehat{\mathbb{Z}}_j|\right] \le cMn^{\theta}2^{j(1-\theta)},$$

where c only depends on ψ and θ .

Proof of Lemma 23. The proof is straightforward when $\theta = 0$ and we assume from now on that $\theta \in (0, 1)$. We deduce from Lemma 19,

(44)
$$\mathbb{E}\left[|\widehat{\mathbb{Z}}_j \cap \overline{\mathbb{Z}}_j|\right] \le |\overline{\mathbb{Z}}_j| \le c_1 M n^{\theta} 2^{j(1-\theta)}$$

We define for $r \ge 0, j \ge -1$,

$$\overline{\mathbb{Z}}_{j,r} = \left\{ k \in \mathbb{Z}, \, 2^{-r-1} < f_{j,k} \le 2^{-r} \right\}$$

and use Lemma 19 to get

$$|\overline{\mathbb{Z}}_{j,r}| \le c_2 M 2^{r\theta} 2^{j(1-\theta)},$$

where c_2 only depends on ψ . Now,

(45)

$$\mathbb{E}\left[\left|\widehat{\mathbb{Z}}_{j} \cap \overline{\mathbb{Z}}_{j}^{c}\right|\right] \leq c_{3}n \sum_{\substack{k \in \mathbb{Z} \\ f_{j,k} \leq 1/n}} f_{j,k} \\ \leq c_{3}n \sum_{2^{r} \geq n} |\overline{\mathbb{Z}}_{j,r}| 2^{-r} \\ \leq c_{4}n M 2^{j(1-\theta)} \sum_{2^{r} \geq n} 2^{-r(1-\theta)} \\ \leq c_{5}M 2^{j(1-\theta)} n^{\theta}.$$

We group (44) and (45) together to end the proof.

Lemma 24. Let $p \in (0,\infty]$, $\alpha \in ((1/p - 1)_+, \tau)$, $\theta \in [0,1) \cap [0,p]$, R > 0, $M \ge 1$ and $f \in \mathscr{F}_{\alpha,p,\theta}(R,M)$. Let $\pi = 1/\min\{1, \min_{x \in \mathbb{R}, \psi(x) \neq 0} \psi^2(x)\}$ and ℓ be an integer such that $2^{\ell} \le M^{1/(1-\theta)} < 2^{\ell+1}$.

1. For all $2^r > \pi n$ such that $r \ge \ell + j + 1$, for some $j \ge -1$, $\widehat{\mathbb{Z}}_{j,r}(\ell) = \emptyset$. 2. If $p \ge 1$, then for all $j \ge 0$, and $r \ge \ell + j$,

(46)
$$\mathbb{E}\left[\widehat{B}_{j,r,\ell}(\emptyset)\right] \leq cRM^{1-1/p}2^{-j(\alpha-\theta/p+\theta)}2^{(1-1/p)r\theta}.$$

3. For all $j \geq -1$, $r = j + \ell$, and subset $K_{j,r,\ell} \subset \widehat{\mathbb{Z}}_{j,r}(\ell) \cap \overline{\mathbb{Z}}_{j}$,

(47)
$$\mathbb{E}\left[\widehat{\mathcal{E}}_{j,r,\ell}(|K_{j,r,\ell}|)\right] \le c \left[\sqrt{\frac{\mathbb{E}[|K_{j,r,\ell}|]}{n}} + \frac{\log((j+2)(r+1)n)}{\sqrt{n}}\right].$$

Moreover,

(48)
$$\mathbb{E}\left[\widehat{\mathcal{E}}_{j,r,\ell}(|\widehat{\mathbb{Z}}_{j,r}(\ell) \cap \overline{\mathbb{Z}}_{j}|)\right] \le c\left[\sqrt{\frac{M^{1/(1-\theta)}2^{j}}{n}} + \frac{\log((j+2)(r+1)n)}{\sqrt{n}}\right]$$

4. For all $j \ge -1$, $r \ge j + \ell + 1$, $K_{j,r,\ell} \subset \widehat{\mathbb{Z}}_{j,r}(\ell) \cap \overline{\mathbb{Z}}_j$, and $\theta \ne 0$,

(49)
$$\widehat{\mathcal{E}}_{j,r,\ell}(|K_{j,r,\ell}|)1_{\mathcal{A}} \leq c \left[|K_{j,r,\ell}| \sqrt{\frac{2^{-r} \log_+^2 \left(M n^{\theta} 2^{j(1-\theta)} / |K_{j,r,\ell}| \right)}{n}} + \frac{\log((j+2)(r+1)n)}{\sqrt{n}} \right],$$

where \mathcal{A} is the event appearing in Theorem 5. Moreover,

(50)
$$\mathbb{E}\left[\widehat{\mathcal{E}}_{j,r,\ell}(|\widehat{\mathbb{Z}}_{j,r}(\ell) \cap \overline{\mathbb{Z}}_{j}|)1_{\mathcal{A}}\right] \leq c \left[M2^{r\theta+j(1-\theta)}\sqrt{\frac{2^{-r}\log_{+}(n2^{-r})}{n}} + \frac{\log((j+2)(r+1)n)}{\sqrt{n}}\right].$$

5. For all $j \ge -1$, $r \ge j + \ell + 1$, and $K_{j,r,\ell} \subset \widehat{\mathbb{Z}}_{j,r}(\ell) \cap \overline{\mathbb{Z}}_{j,r}(\ell)$

(51)
$$\widehat{\mathcal{E}}_{j,r,\ell}(|K_{j,r,\ell}|) \le c \left[|K_{j,r,\ell}| \sqrt{\frac{2^{-r} \log_+^2 (2^r/|K_{j,r,\ell}|)}{n}} + \frac{\log((j+2)(r+1)n)}{\sqrt{n}} \right]$$

Moreover,

(52)
$$\mathbb{E}\left[\widehat{\mathcal{E}}_{j,r,\ell}(|\widehat{\mathbb{Z}}_{j,r}(\ell) \cap \overline{\mathbb{Z}}_{j}|)\right] \leq c \left[M2^{r\theta+j(1-\theta)}\sqrt{\frac{2^{-r}\log_{+}^{2}\left(2^{(r-j)(1-\theta)}/M\right)}{n}} + \frac{\log((j+2)(r+1)n)}{\sqrt{n}}\right].$$

6. For all $j \ge -1$, $r \ge j + \ell + 1$ such that $2^r \le \varsigma_j n / \log((j+2)n)$,

(53)
$$\widehat{\mathcal{E}}_{j,r,\ell}(|\widehat{\mathbb{Z}}_{j,r}(\ell) \cap \overline{\mathbb{Z}}_{j}|) \le c \left[|\widehat{\mathbb{Z}}'_{j,r}| \sqrt{\frac{2^{-r}}{n}} + \frac{\log((j+2)(r+1)n)}{\sqrt{n}} \right]$$

Moreover,

(54)
$$\mathbb{E}[|\widehat{\mathbb{Z}}'_{j,r}|1_{\mathcal{A}}] \le cM2^{r\theta+j(1-\theta)}$$

In all these inequalities, c only depends on ψ , p, θ .

Proof of Lemma 24. The first point is true because ψ is piecewise constant. We now assume that $2^r \leq \pi n$ in the rest of the proof.

We turn to the second point. We have,

$$\widehat{B}_{j,r,\ell}(\emptyset) \le 2^{-j/2} \sum_{k \in \widehat{\mathbb{Z}}_{j,r}(\ell) \cap \overline{\mathbb{Z}}_j} |\beta_{j,k}|.$$

This gives (46) when p = 1 as $f \in \mathcal{B}^{\alpha}_{1,\infty}(R)$. When p > 1, we use (36) to get

$$\widehat{B}_{j,r,\ell}(\emptyset) \leq c_1 2^{-j/2} \|\beta_{j,\cdot}\|_{p,\infty} |\widehat{\mathbb{Z}}_{j,r}(\ell) \cap \overline{\mathbb{Z}}_j|^{1-1/p}.$$

We then take the expectation, apply Jensen's inequality and (42) to get (46).

We now show (47). By using Jensen's inequality,

(55)
$$\mathbb{E}\left[\widehat{\mathcal{E}}_{j,r,\ell}(|K_{j,r,\ell}|)\right] \le c_2 \left[\sqrt{\frac{\mathbb{E}[|K_{j,r,\ell}|]}{n}} + \frac{\mathbb{E}[|K_{j,r,\ell}|]\log_+\left(2^{r+1}/\mathbb{E}[|K_{j,r,\ell}|]\right)}{n} + \sqrt{\frac{\log((j+2)(r+1)n)}{n}} + \frac{\log((j+2)(r+1)n)}{n}\right].$$

Note that $|K_{j,r,\ell}| \leq |\widehat{\mathbb{Z}}_{j,r}(\ell)|$ and

(56)
$$|\widehat{\mathbb{Z}}_{j,r}(\ell)| \le 2^{r+1} \sum_{k \in \mathbb{Z}} \widehat{\sigma}_{j,k}^2 \le c_3 2^r$$

It then follows from the inequality $2^r \leq \pi n$ and from elementary computations that the second term in (55) is not smaller than the first one, up to a multiplicative factor.

As to (48), we remark that (42) becomes

$$\mathbb{E}[|\widehat{\mathbb{Z}}_{j,r}(\ell) \cap \overline{\mathbb{Z}}_j|] \le c_4 M^{1/(1-\theta)} 2^j$$

when $r = j + \ell$. We then use (47) with $K_{j,r,\ell} = \widehat{\mathbb{Z}}_{j,r}(\ell) \cap \overline{\mathbb{Z}}_{j}$.

We now prove (49). Observe that for all $2^r \leq \pi n$, and $r \geq \ell + j + 1$,

$$\widehat{\lambda}_{j,r} \le c_5 \min\left\{ |\widehat{\mathbb{Z}}_j| + \log n, 2^r \right\},\$$

where c_5 only depends on the wavelet basis (this uses $|\widehat{\mathbb{Z}}'_{j,r}| \leq |\widehat{\mathbb{Z}}_j|$ when $2^r \leq \varsigma_j n / \log((j+2)n)$). We deduce from (20) that on \mathcal{A} ,

$$\widehat{\lambda}_{j,r} \le c_6(\mathbb{E}[|\widehat{\mathbb{Z}}_j|] + \log n).$$

As $M \ge 1$, and $\theta > 0$, we deduce from (43),

$$\widehat{\lambda}_{j,r} \le c_7 M n^{\theta} 2^{j(1-\theta)},$$

and using (56),

(57)
$$\widehat{\mathcal{E}}_{j,r,\ell}(|K_{j,r,\ell}|)1_{\mathcal{A}} \leq c_8 |K_{j,r,\ell}| \sqrt{\frac{2^{-r} \log_+ \left(M n^{\theta} 2^{j(1-\theta)} / |K_{j,r,\ell}|\right)}{n}} + c_8 |K_{j,r,\ell}| \frac{\log_+ \left(M n^{\theta} 2^{j(1-\theta)} / |K_{j,r,\ell}|\right)}{n} + c_8 \frac{\log((j+2)(r+1)n)}{\sqrt{n}}.$$

We then use the condition on r to get (49). As to (50), we apply (57) with $K_{j,r,\ell} = \widehat{\mathbb{Z}}_{j,r}(\ell) \cap \overline{\mathbb{Z}}_j$. We take the expectation, apply Jensen's inequality and (42). This yields

$$\mathbb{E}\left[\widehat{\mathcal{E}}_{j,r,\ell}(|\widehat{\mathbb{Z}}_{j,r}(\ell)\cap\overline{\mathbb{Z}}_{j}|)1_{\mathcal{A}}\right] \leq c_{9}M2^{r\theta+j(1-\theta)}\sqrt{\frac{2^{-r}\log_{+}(n2^{-r})}{n}} + c_{9}M2^{r\theta+j(1-\theta)}\frac{\log_{+}(n2^{-r})}{n} + c_{9}\frac{\log((j+2)(r+1)n)}{\sqrt{n}}.$$

We then remark that the second term is not smaller than the first one thanks to the condition on r (up to a multiplicative factor).

The proof of (51) is merely based on the inequality $\widehat{\lambda}_{j,r} \leq 2^{r+1}$, the condition on r, and on (56). The proof of (52) then follows from Jensen's inequality and (42).

We turn to the proof of (53). Here, we use $\widehat{\lambda}_{j,r} \leq |\widehat{\mathbb{Z}}'_{j,r}|$. By doing as in the proof of (56), $|\widehat{\mathbb{Z}}'_{j,r}| \leq c_{10}2^r$. Note also that the maps $x \mapsto x\sqrt{\log_+(a/x)}$ and $x \mapsto x\log_+(a/x)$ are increasing. By using moreover $\widehat{\mathbb{Z}}_{j,r}(\ell) \subset \widehat{\mathbb{Z}}'_{j,r}$,

$$\widehat{\mathcal{E}}_{j,r,\ell}(|\widehat{\mathbb{Z}}_{j,r}(\ell) \cap \overline{\mathbb{Z}}_j|) \mathbf{1}_{\mathcal{A}} \le c_{11} \left[|\widehat{\mathbb{Z}}'_{j,r}| \sqrt{\frac{2^{-r}}{n}} + \frac{|\widehat{\mathbb{Z}}'_{j,r}|}{n} + \frac{\log((j+2)(r+1)n)}{\sqrt{n}} \right].$$

By using the condition on r, the second term in the above inequality is not larger than the first one, up to a multiplicative factor.

Finally, the proof of (54) comes from the embedding $\widehat{\mathbb{Z}}'_{j,r} \subset \mathbb{Z}'_{j,r}$ (see (22)) valid on \mathcal{A} , and from the following inequalities valid for all t > 0,

$$\begin{split} \left| \left\{ k \in \mathbb{Z}, \ \sigma_{j,k}^2 \ge t \right\} \right| &\leq \left| \left\{ k \in \mathbb{Z}, \ f_{j,k} \ge c_{12}t \right\} \right| \\ &\leq c_{13} M^{\theta} t^{-\theta} 2^{j(1-\theta)} \end{split}$$

thanks to Lemma 19.

Lemma 25. Let $p \in (0, \infty]$, $\alpha \in ((1/p - 1)_+, \tau)$, R > 0, and f be a density of $\mathcal{WB}^{\alpha}_{p,\infty}(R)$. Then, there exist C > 0 and q > 1 such that $||f||_q \leq C$. Moreover, C only depends on the wavelet basis, p, α and R.

Proof of Lemma 25. The proof of this lemma is deferred to Appendix C. \Box

4.3. Proof of Theorem 2. Throughout this section, ℓ denotes an integer such that

$$2^{\ell} \leq M^{1/(1-\theta)} < 2^{\ell+1}$$

We now introduce for $K_{j,r,\ell} \subset \widehat{\mathbb{Z}}_{j,r}(\ell) \cap \overline{\mathbb{Z}}_j$,

$$\widehat{T}_{j,r,\ell}(K_{j,r,\ell}) = \left\{ \widehat{B}_{j,r,\ell}(K_{j,r,\ell}) + \widehat{\mathcal{E}}_{j,r,\ell}(|K_{j,r,\ell}|) \right\} \mathbf{1}_{\mathcal{A}},$$

where \mathcal{A} is the event defined in Theorem 5. It follows from Lemmas 21, 24 and 25 that we only need to bound

(58)
$$\sum_{\substack{r \ge \ell - 1\\ 2^r \le \pi n}} \mathbb{E} \left[\inf_{\substack{K_{-1,r,\ell} \subset \widehat{\mathbb{Z}}_{-1,r}(\ell) \cap \overline{\mathbb{Z}}_{-1}}} \widehat{T}_{-1,r,\ell}(K_{-1,r,\ell}) \right] \\ + \sum_{j=0}^{c_0 \log n} \mathbb{E} \left[\inf_{\substack{K_{j,\ell+j,\ell} \subset \widehat{\mathbb{Z}}_{j,\ell+j}(\ell) \cap \overline{\mathbb{Z}}_j}} \widehat{T}_{j,\ell+j,\ell}(K_{j,\ell+j,\ell}) \right] \\ + \sum_{j\ge 0} \sum_{\substack{r \ge \ell+j+1\\ 2^r \le \pi n}} \mathbb{E} \left[\inf_{\substack{K_{j,r,\ell} \subset \widehat{\mathbb{Z}}_{j,r}(\ell) \cap \overline{\mathbb{Z}}_j}} \widehat{T}_{j,r,\ell}(K_{j,r,\ell}) \right] \right]$$

from above (we take $\ell_j = \ell$ for all j). Here, c_0 is a factor depending on α, p, R and the wavelet basis only. We begin by studying the first term of (58).

Lemma 26. Let $p \in (0,\infty]$, $\alpha \in ((1/p-1)_+,\tau)$, $\theta \in [0,1) \cap [0,p]$, R > 0, $M \ge 1$ and $f \in \mathscr{F}_{\alpha,p,\theta}(R,M)$. Then,

(59)
$$\sum_{\substack{r \ge \ell - 1 \\ 2^{T} \le \pi n}} \mathbb{E} \left[\inf_{\substack{K_{-1,r,\ell} \subset \widehat{\mathbb{Z}}_{-1,r}(\ell) \cap \overline{\mathbb{Z}}_{-1}}} \widehat{T}_{-1,r,\ell}(K_{-1,r,\ell}) \right] \\ \le c \left[Mn^{-(1-\theta)} \left[1_{\theta > 1/2} + (\log n) 1_{\theta = 1/2} \right] + M^{1/(2(1-\theta))} n^{-1/2} + (\log^2 n) n^{-1/2} \right],$$

where c only depends on ψ, θ, p .

Proof of Lemma 26. Let A be the left-hand side of (59). We take $K_{-1,r,\ell} = \widehat{\mathbb{Z}}_{-1,r}(\ell) \cap \overline{\mathbb{Z}}_{-1}$ and deduce from (48), and (52) that if $\theta \in [0, 1/2)$,

$$A \leq c_1 \left[\sqrt{\frac{M^{1/(1-\theta)}}{n}} + \frac{\log(\ell n)}{\sqrt{n}} \right]$$
$$+ c_1 \sum_{\substack{r \geq \ell \\ 2^r \leq \pi n}} \left[M 2^{r(\theta-1/2)} \sqrt{\frac{\log_+ \left(n2^{r(1-\theta)}/M\right)}{n}} + \frac{\log((r+1)n)}{\sqrt{n}} \right].$$

We conclude by applying Lemma 30 in Appendix B. If $\theta \in (1/2, 1)$, we rather use (48) and (50) to get

$$A \le c_2 \left[\sqrt{\frac{M^{1/(1-\theta)}}{n}} + \frac{\log(\ell n)}{\sqrt{n}} \right] + c_2 \sum_{\substack{r \ge \ell\\2^r \le \pi n}} \left[M 2^{r(\theta-1/2)} \sqrt{\frac{\log_+ (n2^{-r})}{n}} + \frac{\log((r+1)n)}{\sqrt{n}} \right],$$

and we apply Lemma 30 again. For $\theta = 1/2$, we use (53) and (54), to get with $K_{-1,r,\ell} = \widehat{\mathbb{Z}}_{-1,r}(\ell) \cap \overline{\mathbb{Z}}_{-1}$,

$$\sum_{\substack{r \ge \ell \\ 2^r \le \varsigma_{-1} \pi n / \log n}} \mathbb{E}\left[\widehat{T}_{-1,r,\ell}(K_{-1,r,\ell})\right] \le c_3 \left[M(\log n)n^{-1/2} + (\log^2 n)n^{-1/2}\right].$$

When r is higher, we choose $K_{-1,r,\ell} = \emptyset$ and note:

$$\sum_{\substack{r \ge \ell\\2^r \ge \varsigma_{-1}\pi n/\log n}} \mathbb{E}\left[\widehat{T}_{-1,r,\ell}(K_{-1,r,\ell})\right] \le 2^{1/2} \sum_{k \in \mathbb{Z}} |\beta_{-1,k}| \mathbb{P}\left[\widehat{\sigma}_{-1,k}^2 \le (\log n)/(\varsigma_{-1}n) \cap \mathcal{A}\right]$$
$$\le 2^{1/2} \sum_{\substack{k \in \mathbb{Z}\\\sigma_{-1,k}^2 \le c_4 \log n/n}} |\beta_{-1,k}|.$$

Therefore, the integers k are such that $f_{-1,k} \leq c_5 \log n/n$ and we conclude as in the proof of Lemma 20.

The lemma below deals with the second term of (58).

Lemma 27. Let $p \in (0,\infty]$, $\alpha \in ((1/p-1)_+, \tau)$, $\theta \in [0,1) \cap [0,p]$, R > 0, $M \ge 1$ and $f \in \mathscr{F}_{\alpha,p,\theta}(R,M)$. Then,

(60)

$$\sum_{j=0}^{c_0 \log n} \mathbb{E} \left[\inf_{K_{j,\ell+j,\ell} \subset \widehat{\mathbb{Z}}_{j,\ell+j}(\ell) \cap \overline{\mathbb{Z}}_j} \widehat{T}_{j,\ell+j,\ell}(K_{j,\ell+j,\ell}) \right] \\
\leq c_1 \left[R^{1/(1+2\alpha)} M^{(\alpha-1/p+1)/((1-\theta)(2\alpha+1))} n^{-\alpha/(2\alpha+1)} + M^{1/(2(1-\theta))} n^{-1/2} \right] + c_2 (\log^2 n) n^{-1/2},$$

where c_1 only depends on α, p, θ, ψ , and where c_2 only depends on $\alpha, p, \theta, \psi, R$.

Proof of Lemma 27. We first suppose that $p \ge 1$. We have,

$$\sum_{j=0}^{c_0 \log n} \mathbb{E} \left[\inf_{K_{j,\ell+j,\ell} \subset \widehat{\mathbb{Z}}_{j,\ell+j}(\ell) \cap \overline{\mathbb{Z}}_j} \widehat{T}_{j,\ell+j,\ell}(K_{j,\ell+j,\ell}) \right] \\ \leq \sum_{j=0}^{c_0 \log n} \min \left\{ \mathbb{E} \left[\widehat{T}_{j,\ell+j,\ell}(\emptyset) \right], \mathbb{E} \left[\widehat{T}_{j,\ell+j,\ell}(\widehat{\mathbb{Z}}_{j,\ell+j}(\ell) \cap \overline{\mathbb{Z}}_j) \right] \right\}.$$

We use (46) and (48) to get

$$\sum_{j=0}^{c_0 \log n} \mathbb{E} \left[\inf_{K_{j,\ell+j,\ell} \subset \widehat{\mathbb{Z}}_{j,\ell+j}(\ell) \cap \overline{\mathbb{Z}}_j} \widehat{T}_{j,\ell+j,\ell}(K_{j,\ell+j,\ell}) \right] \\ \leq c_1 \sum_{j=0}^{\infty} \min \left\{ R M^{(1-1/p)/(1-\theta)} 2^{-j\alpha}, \sqrt{\frac{M^{1/(1-\theta)} 2^j}{n}} \right\} + c_2 \frac{\log^2 n}{\sqrt{n}} \\ \leq c_3 \left[R^{1/(1+2\alpha)} M^{(\alpha+1-1/p)/((1-\theta)(2\alpha+1))} n^{-\alpha/(2\alpha+1)} + M^{1/(2(1-\theta))} n^{-1/2} \right] \\ + c_2 (\log^2 n) n^{-1/2}.$$

We now suppose p < 1 and consider $j_0 \ge 0$. We deduce from (48),

$$\sum_{j=0}^{j_0} \mathbb{E}\left[\widehat{T}_{j,\ell+j,\ell}(\widehat{\mathbb{Z}}_{j,\ell+j}(\ell) \cap \overline{\mathbb{Z}}_j)\right] \le c_4 \sqrt{M^{1/(1-\theta)} \frac{2^{j_0}}{n}} + c_5 \frac{\log^2 n}{\sqrt{n}}$$

We moreover set $\eta_j^{1/p-1/2} = R^{-1/2} 2^{(j/2)(\alpha+1/2)} n^{-1/(2p)}$ and

$$K_{j,\ell+j,\ell} = \left\{ k \in \widehat{\mathbb{Z}}_{j,\ell+j}(\ell) \cap \overline{\mathbb{Z}}_j, \, |\beta_{j,k}| \geq \eta_j \right\}.$$

As $f \in \mathcal{WB}^{\alpha}_{p,\infty}(R)$,

$$|K_{j,\ell+j,\ell}| \le \eta_j^{-p} R^p 2^{-jp(\alpha+1/2-1/p)} \le R^{2/(2/p-1)} 2^{-2j(\alpha+1-1/p)/(2/p-1)} n^{1/(2/p-1)}.$$

Moreover, (35) leads to

$$\sum_{k \notin K_{j,\ell+j,\ell}} 2^{-j/2} |\beta_{j,k}| \le c_6 \eta_j^{1-p} R^p 2^{-jp(\alpha+1/2-1/(2p))} \le c_6 R^{1/(2/p-1)} 2^{-j(\alpha+1-1/p)/(2/p-1)} n^{-(1/p-1)/(2/p-1)}.$$

Therefore, (47) gives

$$\mathbb{E}\left[\widehat{T}_{j,\ell+j,\ell}(K_{j,\ell+j,\ell})\right] \le c_7 \left[n^{-(1/p-1)/(2/p-1)} R^{1/(2/p-1)} 2^{-j(\alpha+1-1/p)/(2/p-1)} + \frac{\log n}{\sqrt{n}} \right]$$

and hence

$$\sum_{j=j_0}^{c_0 \log n} \mathbb{E}\left[\widehat{T}_{j,\ell+j,\ell}(K_{j,\ell+j,\ell})\right] \le c_8 n^{-(1/p-1)/(2/p-1)} R^{1/(2/p-1)} 2^{-j_0(\alpha+1-1/p)/(2/p-1)} + c_9 \frac{\log^2 n}{\sqrt{n}}.$$

We conclude by choosing j_0 appropriately.

It then remains to deal with the last term in (58), namely,

(61)
$$\sum_{j\geq 0} \sum_{\substack{r\geq \ell+j+1\\ 2^r\leq \pi n}} \mathbb{E}\left[\inf_{K_{j,r,\ell}\subset \widehat{\mathbb{Z}}_{j,r}(\ell)\cap \overline{\mathbb{Z}}_j} \widehat{T}_{j,r,\ell}(K_{j,r,\ell})\right].$$

We distinguish different cases in the sections below.

4.4. Proof of Theorem 2 when $p \ge 1$ and $\theta \ne \alpha/(2\alpha + 1 - 1/p)$. To make the proof more concise, we assume temporarily that $\theta \ne 0$. We deduce from (46), (50), (52) that for c_1 large enough

$$-\frac{\log^3 n}{\sqrt{n}} + \frac{1}{c_1} \sum_{\substack{j=0\\2^r \le \pi n}}^{\infty} \sum_{\substack{r \ge \ell+j+1\\2^r \le \pi n}} \min\left\{ \mathbb{E}\left[\widehat{T}_{j,r,\ell}(\emptyset)\right], \mathbb{E}\left[\widehat{T}_{j,r,\ell}(\widehat{\mathbb{Z}}_{j,r}(\ell) \cap \overline{\mathbb{Z}}_j)\right] \right\}$$
$$\leq \sum_{\substack{j=0\\2^r \ge \ell+j+1\\2^r \le \pi n}}^{\infty} \min\left\{ RM^{1-1/p} 2^{-j(\alpha-\theta/p+\theta)} 2^{(1-1/p)r\theta}, M2^{r\theta+j(1-\theta)} \sqrt{\frac{2^{-r}\log_+\left(n2^{-r}\right)}{n}}, M2^{r\theta+j(1-\theta)} \sqrt{\frac{2^{-r}\log_+\left(2^{(r-j)(1-\theta)}/M\right)}{n}} \right\}.$$

Hence,

$$-\frac{\log^{3} n}{\sqrt{n}} + \frac{1}{c_{1}} \sum_{j=0}^{\infty} \sum_{\substack{r \ge \ell+j+1 \\ 2^{r} \le \pi n}} \min\left\{ \mathbb{E}\left[\widehat{T}_{j,r,\ell}(\emptyset)\right], \mathbb{E}\left[\widehat{T}_{j,r,\ell}(\widehat{\mathbb{Z}}_{j,r}(\ell) \cap \overline{\mathbb{Z}}_{j})\right] \right\}$$

$$\leq \sum_{r \ge 1} \sum_{\substack{j \ge 0 \\ 2^{j} \le \pi n 2^{1-r} M^{-1/(1-\theta)}}} \min\left\{ RM^{(1-1/p)/(1-\theta)} 2^{-j\alpha} 2^{(1-1/p)r\theta}, M^{1/(2(1-\theta))} 2^{r(\theta-1/2)+j/2} \frac{\sqrt{\log_{+}\left(n 2^{-r-j} M^{-1/(1-\theta)}\right)}}{\sqrt{n}}, M^{1/(2(1-\theta))} 2^{r(\theta-1/2)+j/2} \frac{r}{\sqrt{n}} \right\}.$$

This expression is not larger than $B_1 + B_2$ where

$$B_{1} = RM^{(1-1/p)/(1-\theta)} \sum_{r \ge 1} \sum_{\substack{j \ge 0 \\ 2^{j} \le \pi n 2^{1-r}M^{-1/(1-\theta)} \\ \pi R^{2}M^{-(2/p-1)/(1-\theta)}n 2^{r(1-2\theta/p)} \le 2^{j(1+2\alpha)}}} 2^{-j\alpha} 2^{(1-1/p)r\theta}$$

$$B_{2} = M^{1/(2(1-\theta))} \sum_{r \ge 1} \sum_{\substack{j \ge 0 \\ 2^{j} \le \pi n 2^{1-r}M^{-1/(1-\theta)} \\ \pi R^{2}M^{-(2/p-1)/(1-\theta)}n 2^{r(1-2\theta/p)} > 2^{j(1+2\alpha)}}} \left\{ 2^{r(\theta-1/2)+j/2} \frac{\sqrt{\log_{+} \left(n 2^{-r-j}M^{-1/(1-\theta)}\right)}}{\sqrt{n}}, 2^{r(\theta-1/2)+j/2} \frac{r}{\sqrt{n}} \right\}.$$

Note that if $2^j \leq \pi n 2^{1-r} M^{-1/(1-\theta)}$ and if $\pi R^2 M^{-(2/p-1)/(1-\theta)} n 2^{r(1-2\theta/p)} \leq 2^{j(1+2\alpha)}$, then

$$2^{j\alpha} \ge 2^{-1/2} R M^{(1-1/p)/(1-\theta)} 2^{r(1-\theta/p)}.$$

We deduce,

$$B_{1} \leq c_{2}RM^{(1-1/p)/(1-\theta)} \sum_{r\geq 1} 2^{(1-1/p)r\theta} \min\left\{R^{-1}M^{-(1-1/p)/(1-\theta)}2^{-r(1-\theta/p)}, \\ \left[R^{2}M^{-(2/p-1)/(1-\theta)}n2^{r(1-2\theta/p)}\right]^{-\alpha/(2\alpha+1)}\right\}.$$

This sum is not larger than $c_2[B_{1,1} + B_{1,2}]$, where

$$B_{1,1} = \sum_{r \ge r_0} 2^{-r(1-\theta)}$$
$$B_{1,2} = \sum_{r < r_0} a_r$$

where

(62)
$$a_r = R^{1/(2\alpha+1)} M^{(\alpha+1-1/p)/((2\alpha+1)(1-\theta))} 2^{r \frac{2\alpha+1-1/p}{2\alpha+1} \left[\theta - \frac{\alpha}{2\alpha+1-1/p}\right]} n^{-\alpha/(2\alpha+1)},$$

and where $r_0 \ge 1$ is the smallest integer such that

$$2^{r_0(\alpha+1-\theta/p)} > R^{-1}M^{-(\alpha+1-1/p)/(1-\theta)}n^{\alpha}.$$

Note that

$$B_{1,1} \le c_3 2^{-r_0(1-\theta)} \le c_3 R^{(1-\theta)/(1+\alpha-\theta/p)} M^{(\alpha+1-1/p)/(\alpha+1-\theta/p)} n^{-\alpha(1-\theta)/(\alpha+1-\theta/p)}.$$

As to $B_{1,2}$, we remark that $B_{1,2} = 0$ if $r_0 = 1$. We may therefore assume that $r_0 \ge 2$, which implies the reverse inequality

$$2^{(r_0-1)(\alpha+1-\theta/p)} \leq R^{-1} M^{-(\alpha+1-1/p)/(1-\theta)} n^{\alpha}.$$

Note now that the exponent in the sum is negative when $\theta < \alpha/(2\alpha + 1 - 1/p)$, and hence

$$B_{1,2} \le c_4 a_1$$

$$\le c_5 R^{1/(2\alpha+1)} M^{(\alpha+1-1/p)/((2\alpha+1)(1-\theta))} n^{-\alpha/(2\alpha+1)}.$$

When $\theta > \alpha/(2\alpha + 1 - 1/p)$, the exponent is positive and

$$B_{1,2} \le c_6 a_{r_0} \le c_7 R^{(1-\theta)/(1+\alpha-\theta/p)} M^{(\alpha+1-1/p)/(\alpha+1-\theta/p)} n^{-\alpha(1-\theta)/(\alpha+1-\theta/p)}.$$

By gathering these results, we obtain the desired bound on B_1 .

We now deal with B_2 . Lemma 30 in Appendix B gives

$$B_{2} \leq c_{8} \sum_{r \geq 1} \min \left\{ 2^{-r(1-\theta)}, \\ M^{1/(2(1-\theta))} 2^{r(\theta-1/2)} \left(R^{2} M^{-(2/p-1)/(1-\theta)} n 2^{r(1-2\theta/p)} \right)^{1/2/(2\alpha+1)} \times \frac{\sqrt{\log_{+} \left(n 2^{-r} (R^{2} M^{-(2/p-1)/(1-\theta)} n 2^{r(1-2\theta/p)})^{-1/(2\alpha+1)} M^{-1/(1-\theta)} \right)}}{\sqrt{n}}, \\ M^{1/(2(1-\theta))} 2^{r(\theta-1/2)} \left(R^{2} M^{-(2/p-1)/(1-\theta)} n 2^{r(1-2\theta/p)} \right)^{1/2/(2\alpha+1)} \frac{r}{\sqrt{n}} \right\} \\ \leq c_{9} \sum_{r \geq 1} \min \left\{ 2^{-r(1-\theta)}, \ a_{r} \sqrt{\log_{+} C_{r}}, \ a_{r} r \right\} \\ \leq c_{9} \left[B_{2,1} + B_{2,2} \right]$$

where

(63)
$$C_r = R^{-2/(2\alpha+1)} M^{-2(\alpha+1-1/p)/((2\alpha+1)(1-\theta))} n^{2\alpha/(2\alpha+1)} 2^{-2r(\alpha+1-\theta/p)/(2\alpha+1)}$$

and where $B_{2,1}$ and $B_{2,2}$ are defined when $\theta < \alpha/(2\alpha + 1 - 1/p)$ by

$$B_{2,1} = \sum_{r=1}^{\infty} ra_r$$
$$B_{2,2} = 0.$$

When $\theta > \alpha/(2\alpha + 1 - 1/p)$, we rather set

$$B_{2,1} = \sum_{r \ge r_0} 2^{-r(1-\theta)}$$

and

$$B_{2,2} = \sum_{r < r_0} a_r \sqrt{\log_+ C_r}$$

When $\theta < \alpha/(2\alpha + 1 - 1/p)$,

$$B_{2,1} \le c_{10}a_1$$

$$\le c_{11}R^{1/(2\alpha+1)}M^{(\alpha+1-1/p)/((2\alpha+1)(1-\theta))}n^{-\alpha/(2\alpha+1)}.$$

When $\theta > \alpha/(2\alpha + 1 - 1/p)$, the sum $B_{2,1}$ is equal to $B_{1,1}$ and has already been bounded, see the above. Moreover, we deduce from Lemma 30,

$$B_{2,2} \le c_{12} a_{r_0} \sqrt{\log_+(C_{r_0})} \le c_{13} R^{(1-\theta)/(1+\alpha-\theta/p)} M^{(\alpha+1-1/p)/(\alpha+1-\theta/p)} n^{-\alpha(1-\theta)/(\alpha+1-\theta/p)}$$

It then remains to group all these results to obtain the wished bound on B_2 .

Let us now remark that the condition $\theta \neq 0$ was made in the proof in order to use (50) when $\theta > \alpha/(2\alpha + 1 - 1/p)$. It is not necessary when $\theta < \alpha/(2\alpha + 1 - 1/p)$, which is obviously the case when $\theta = 0$. The proof remains therefore valid when $\theta = 0$.

4.5. Proof of Theorem 2 when p > 1 and $\theta = \alpha/(2\alpha + 1 - 1/p)$. The previous proof can easily be adapted to deal with $\theta = \alpha/(2\alpha + 1 - 1/p)$. However, it would lead to additional logarithmic factors. A slight refinement can be obtained thanks to (53). We deduce from Lemma 24 that for c_1 large enough,

$$-\frac{\log^{3} n}{\sqrt{n}} + \frac{1}{c_{1}} \sum_{j=0}^{\infty} \sum_{\substack{\substack{r \ge \ell+j+1\\2^{r} \le \pi n\\2^{r} \le qn}} \min\left\{\mathbb{E}\left[\widehat{T}_{j,r,\ell}(\emptyset)\right], \mathbb{E}\left[\widehat{T}_{j,r,\ell}(\widehat{\mathbb{Z}}_{j,r}(\ell) \cap \overline{\mathbb{Z}}_{j})\right]\right\}$$
$$\leq \sum_{\substack{r \ge 1\\2^{j} \le \pi n 2^{1-r}M^{-1/(1-\theta)}\\2^{j} < 2\varsigma_{i}(n/\log n)2^{-r}M^{-1/(1-\theta)}}} \left\{RM^{(1-1/p)/(1-\theta)}2^{-j\alpha}2^{(1-1/p)r\theta}, \frac{M^{1/(2(1-\theta))}2^{r(\theta-1/2)+j/2}}{\sqrt{n}}\right\}$$

We may now bound the right-hand side of this inequality by $B_1 + B_2$ where B_1 has been defined in the preceding section by

$$B_{1} = RM^{(1-1/p)/(1-\theta)} \sum_{r \ge 1} \sum_{\substack{j \ge 0\\ 2^{j} \le \pi n 2^{1-r} M^{-1/(1-\theta)}\\ R^{2}M^{-(2/p-1)/(1-\theta)} n 2^{r(1-2\theta/p)} \le 2^{j(1+2\alpha)}}} 2^{-j\alpha} 2^{(1-1/p)r\theta}$$

and where

$$B_2 = M^{1/(2(1-\theta))} \sum_{r \ge 1} \sum_{\substack{j \ge 0\\ 2^j \le \pi n 2^{1-r} M^{-1/(1-\theta)}\\ R^2 M^{-(2/p-1)/(1-\theta)} n 2^{r(1-2\theta/p)} > 2^{j(1+2\alpha)}} \frac{2^{r(\theta-1/2)+j/2}}{\sqrt{n}}$$

corresponds to the definition of the preceding section, up to log factors. Similar computations then lead to

$$B_1 + B_2 \le c_2 R^{1/(2\alpha+1)} M^{(\alpha+1-1/p)/((2\alpha+1)(1-\theta))} n^{-\alpha/(2\alpha+1)} \log n.$$

It then remains to deal with

$$\begin{split} \sum_{j\geq 0} \sum_{\substack{r\geq \ell+j+1\\ 2^r\leq \pi n\\ 2^r>\varsigma_j n/\log((j+2)n)}} \mathbb{E} \left[\inf_{\substack{K_{j,r,\ell}\subset \widehat{\mathbb{Z}}_{j,r}(\ell)\cap \overline{\mathbb{Z}}_j}} \widehat{T}_{j,r,\ell}(K_{j,r,\ell}) \right] \\ &\leq \sum_{j\geq 0} \sum_{\substack{r\geq \ell+j+1\\ 2^r>\varsigma_j n/\log n}} \mathbb{E} \left[\widehat{T}_{j,r,\ell}(\emptyset) \right] \\ &\leq \sum_{j\geq 0} 2^{-j/2} \sum_{k\in \mathbb{Z}} |\beta_{j,k}| \mathbb{P} \left[\widehat{\sigma}_{j,k}^2 \leq (\log n)/(\varsigma_j n) \cap \mathcal{A} \right] \\ &\leq \sum_{j\geq 0} 2^{-j/2} \sum_{\substack{k\in \mathbb{Z}\\ \sigma_{j,k}^2 \leq c_3 \log n/n}} |\beta_{j,k}| \\ &\leq \sum_{j\geq 0} 2^{-j/2} \sum_{\substack{k\in \mathbb{Z}\\ f_{j,k} \leq c_4 \log n/n}} |\beta_{j,k}|. \end{split}$$

We conclude by applying Lemma 20.

4.6. Proof of Theorem 2 when p = 1 and $\theta = \alpha/(2\alpha + 1 - 1/p)$. Since $f \in \mathcal{B}^{\alpha}_{1,\infty}(R)$,

(64)

$$\sum_{r \ge \ell+j+1} \widehat{B}_{j,r,\ell}(\emptyset) \le 2^{-j/2} \sum_{r \ge \ell+j+1} \sum_{k \in \widehat{\mathbb{Z}}_{j,r}(\ell) \cap \overline{\mathbb{Z}}_j} |\beta_{j,k}|$$

$$\le R2^{-j\alpha}.$$

Let us remark that the boundary $\theta = \alpha/(2\alpha + 1 - 1/p)$ is $\theta = 1/2$ here. We deduce from (53) and (54) that if $2^r \leq \varsigma_j n/\log((j+2)n)$,

$$\mathbb{E}\left[\widehat{\mathcal{E}}_{j,r,\ell}(|\widehat{\mathbb{Z}}_{j,r}(\ell)\cap\overline{\mathbb{Z}}_{j}|)1_{\mathcal{A}}\right] \leq c_{1}\left[\frac{M2^{j/2}}{\sqrt{n}} + \frac{\log((j+2)(r+1)n)}{\sqrt{n}}\right].$$

Therefore,

(65)
$$\sum_{\substack{r \ge \ell+j+1\\ 2^r \le \varsigma_j n/\log((j+2)n)}} \mathbb{E}\left[\widehat{\mathcal{E}}_{j,r,\ell}(|\widehat{\mathbb{Z}}_{j,r}(\ell) \cap \overline{\mathbb{Z}}_j|) \mathbf{1}_{\mathcal{A}}\right] \le c_2 \left[\frac{M2^{j/2}\log n}{\sqrt{n}} + \frac{\log^2 n}{n}\right].$$

We put (65) and (64) together and conclude the proof as in the preceding section.

4.7. Proof of Theorem 2 when p < 1 and $\theta \neq \alpha/(2\alpha + 1 - 1/p)$. As in Section 4.4, we first suppose that $\theta \neq 0$. We define for $r \geq \ell + j + 1$,

$$K_{j,r,\ell} = \left\{ k \in \widehat{\mathbb{Z}}_{j,r}(\ell) \cap \overline{\mathbb{Z}}_j, \, |\beta_{j,k}| \geq 2^{-(r-j)/2} n^{-1/2} \right\}.$$

As $f \in \mathcal{WB}_{p,\infty}^{\alpha}(R)$ and (35), we get $|K_{j,r,\ell}| \leq k_{j,r}$ and

(66)
$$\sum_{k \notin K_{j,r,\ell}} 2^{-j/2} |\beta_{j,k}| \le c_1 k_{j,r} 2^{-r/2} n^{-1/2}$$

where $k_{j,r} = R^p n^{p/2} 2^{-jp(\alpha+1-1/p)} 2^{rp/2}$.

We consider r and j such that $2^r \leq \pi n$, and $j \leq r - \ell - 1$. We deduce from (49), (51), and the conditions on r, j,

$$\widehat{T}_{j,r,\ell}(K_{j,r,\ell}) \le c_2 k_{j,r} 2^{-r/2} n^{-1/2} \log_+ \left(M n^{\theta - p/2} 2^{-rp/2} 2^{jp(\alpha + 1 - \theta/p)} / R^p \right) + c_3 (\log n) n^{-1/2}$$

$$\widehat{T}_{j,r,\ell}(K_{j,r,\ell}) \le c_4 k_{j,r} 2^{-r/2} n^{-1/2} \log_+ \left(n^{-p/2} 2^{r(1-p/2)} 2^{jp(\alpha + 1 - 1/p)} / R^p \right) + c_5 (\log n) n^{-1/2}.$$

We derive from these two inequalities, from (50) and (52),

$$-\frac{\log^{3} n}{\sqrt{n}} + \frac{1}{c_{6}} \sum_{j \ge 0} \sum_{\substack{r \ge \ell+j+1\\2^{r} \le \pi n}} \mathbb{E} \left[\inf_{\substack{K_{j,r,\ell} \subset \widehat{\mathbb{Z}}_{j,r}(\ell) \cap \overline{\mathbb{Z}}_{j}}} \widehat{T}_{j,r,\ell}(K_{j,r,\ell}) \right]$$
$$\leq \sum_{r \ge 1} \sum_{\substack{j \ge 0\\2^{j} \le \pi n 2^{1-r} M^{-1/(1-\theta)}}} \min\{A_{j,r}, A'_{j,r}, B_{j,r}, B'_{j,r}\}$$

where

$$\begin{aligned} A_{j,r} &= R^p M^{-(1-p)/(2(1-\theta))} n^{-(1-p)/2} 2^{-r(1-p)/2} 2^{-jp(\alpha-1/(2p)+1/2)} \\ &\times \log_+ \left(M^{1-p/(2(1-\theta))} n^{\theta-p/2} 2^{-rp/2} 2^{jp(\alpha+1/2-\theta/p)} / R^p \right) \\ A'_{j,r} &= R^p M^{-(1-p)/(2(1-\theta))} n^{-(1-p)/2} 2^{-r(1-p)/2} 2^{-jp(\alpha-1/(2p)+1/2)} \\ &\times \log_+ \left(M^{(1-p/2)/(1-\theta)} n^{-p/2} 2^{r(1-p/2)} 2^{jp(\alpha+1/2)} / R^p \right) \\ B_{j,r} &= M^{1/(2(1-\theta))} 2^{r(\theta-1/2)+j/2} \sqrt{\frac{\log_+ \left(n2^{-r-j} M^{-1/(1-\theta)} \right)}{n}} \\ B'_{j,r} &= M^{1/(2(1-\theta))} 2^{r(\theta-1/2)+j/2} \frac{r}{\sqrt{n}}. \end{aligned}$$

Let j_r be the smallest (possibly negative) integer such that

$$R^2 M^{-(2/p-1)/(1-\theta)} n 2^{-r(2\theta/p-1)} \le 2^{j_r(2\alpha+1)}.$$

Lemma 30 entails when $j_r \ge 0$:

$$\sum_{\substack{j \ge 0\\2^{j} \le \pi n 2^{1-r} M^{-1/(1-\theta)}}} \min\{A_{j,r}, B_{j,r}\}$$

$$\le c_7 \left\{ R^p M^{-(1-p)/(2(1-\theta))} n^{-(1-p)/2} 2^{-r(1-p)/2} 2^{-j_r p(\alpha-1/(2p)+1/2)} \\ \times \log_+ \left(M^{1-p/(2(1-\theta))} n^{\theta-p/2} 2^{-rp/2} 2^{j_r p(\alpha+1/2-\theta/p)} / R^p \right) \\ + M^{1/(2(1-\theta))} 2^{r(\theta-1/2)+j_r/2} \sqrt{\frac{\log_+ \left(n 2^{-r-j_r} M^{-1/(1-\theta)} \right)}{n}} \right\}$$

$$\le c_8 a_r \log_+ C_r,$$

where a_r and C_r have been defined in Section 4.4 by (62) and (63). The same results holds true when $j_r < 0$ since then

$$\sum_{\substack{j \ge 0\\2^{j} \le \pi n 2^{1-r} M^{-1/(1-\theta)}}} \min\{A_{j,r}, B_{j,r}\} \le \sum_{j \ge 0} A_{j+j_r,r}$$
$$\le c_9 R^p M^{-(1-p)/(2(1-\theta))} n^{-(1-p)/2} 2^{-r(1-p)/2} 2^{-j_r p(\alpha-1/(2p)+1/2)} \times \log_+ \left(M^{1-p/(2(1-\theta))} n^{\theta-p/2} 2^{-rp/2} 2^{j_r p(\alpha+1/2-\theta/p)} / R^p \right)$$
$$\le c_{10} a_r \log_+ C_r.$$

Likewise, by supposing without loss of generality that $j_r \ge 0$,

$$\sum_{\substack{j \ge 0\\2^{j} \le \pi n 2^{1-r} M^{-1/(1-\theta)}}} \min\{A'_{j,r}, B'_{j,r}\}$$
$$\le c_{11} \left\{ R^{p} M^{-(1-p)/(2(1-\theta))} n^{-(1-p)/2} 2^{-r(1-p)/2} 2^{-j_{r}p(\alpha-1/(2p)+1/2)} \right. \\ \left. \times \log_{+} \left(M^{(1-p/2)/(1-\theta)} n^{-p/2} 2^{r(1-p/2)} 2^{j_{r}p(\alpha+1/2)} / R^{p} \right) \right\}$$

$$+M^{1/(2(1-\theta))}2^{r(\theta-1/2)+j_r/2}\frac{r}{\sqrt{n}}$$

$$\leq c_{12}ra_r.$$

We also have,

$$\sum_{\substack{j \ge 0\\2^j \le \pi n 2^{1-r} M^{-1/(1-\theta)}}} B_{j,r} \le c_{13} 2^{-r(1-\theta)}$$

Therefore,

$$\sum_{r\geq 1} \sum_{\substack{j\geq 0\\2^{j}\leq \pi n 2^{1-r}M^{-1/(1-\theta)}}} \min\{A_{j,r}, A'_{j,r}, B_{j,r}, B'_{j,r}\} \leq c_{14} \sum_{r=1}^{\infty} \min\left\{a_r \log_+ C_r, ra_r, 2^{-r(1-\theta)}\right\}.$$

We conclude the proof as in the end of Section 4.4 (by noticing, as previously, that the reasoning does not need (49) and (50) when $\theta < \alpha/(2\alpha + 1 - 1/p)$ and is therefore also valid when $\theta = 0$).

4.8. Proof of Theorem 2 when p < 1 and $\theta = \alpha/(2\alpha + 1 - 1/p)$. We define the same set $K_{j,r,\ell}$ and the same real number $k_{j,r}$ as in the preceding section. When $2^r \leq \varsigma_j n/\log((j+2)n)$, $\hat{\lambda}_{j,r} \leq |\hat{\mathbb{Z}}'_{j,r}|$ and hence

$$\begin{aligned} \widehat{\mathcal{E}}_{j,r,\ell}(|K_{j,r,\ell}|) - \frac{\log((j+2)(r+1)n)}{\sqrt{n}} &\leq c_1 k_{j,r} \left[\sqrt{\frac{2^{-r} \log_+ \left(|\widehat{\mathbb{Z}}'_{j,r}|/k_{j,r} \right)}{n}} + \frac{\log_+ \left(|\widehat{\mathbb{Z}}'_{j,r}|/k_{j,r} \right)}{n} \right] \\ &\leq c_2 k_{j,r} \sqrt{\frac{2^{-r} \log_+^2 \left(|\widehat{\mathbb{Z}}'_{j,r}|/k_{j,r} \right)}{n}}. \end{aligned}$$

Jensen's inequality, (54) and (66) imply

$$\mathbb{E}\left[\widehat{T}_{j,r,\ell}(K_{j,r,\ell})\right] - \frac{\log((j+2)(r+1)n)}{\sqrt{n}} \le c_3 k_{j,r} \sqrt{\frac{2^{-r}\log_+^2\left(MR^{-p}2^{r(\theta-p/2)}n^{-p/2}2^{jp(\alpha+1-\theta/p)}\right)}{n}}$$

Note that (53) and (54) entail

$$\mathbb{E}\left[\widehat{T}_{j,r,\ell}(\widehat{\mathbb{Z}}_{j,r}(\ell)\cap\overline{\mathbb{Z}}_{j})\right] \leq c_4\left[\frac{M2^{r(\theta-1/2)+j(1-\theta)}}{\sqrt{n}} + \frac{\log((j+2)(r+1)n)}{\sqrt{n}}\right]$$

We deduce,

$$-\frac{\log^3 n}{\sqrt{n}} + \frac{1}{c_5} \sum_{j \ge 0} \sum_{\substack{r \ge \ell+j+1\\ 2^r \le \pi n\\ 2^r \le \varsigma_j n/\log((j+2)n)}} \mathbb{E}\left[\inf_{\substack{K_{j,r,\ell} \subset \widehat{\mathbb{Z}}_{j,r}(\ell) \cap \overline{\mathbb{Z}}_j}} \widehat{T}_{j,r,\ell}(K_{j,r,\ell})\right]$$
$$\leq \sum_{r \ge 1} \sum_{\substack{2^j \le \pi n 2^{1-r}M^{-1/(1-\theta)}}} \min\{A_{j,r}, B_{j,r}\},$$

where

$$A_{j,r} = R^p M^{-(1-p)/(2(1-\theta))} n^{-(1-p)/2} 2^{-r(1-p)/2} 2^{-jp(\alpha-1/(2p)+1/2)} \\ \times \log_+ \left(M^{(1-p/2)/(1-\theta)} R^{-p} 2^{r(\theta-p/2)} n^{-p/2} 2^{jp(\alpha+1/2)} \right) \\ B_{j,r} = M^{1/(2(1-\theta))} 2^{r(\theta-1/2)+j/2} \sqrt{1/n}.$$

By doing as in the preceding section,

$$\sum_{\substack{j\geq 0\\2^{j}\leq \pi n 2^{1-r}M^{-1/(1-\theta)}}} \min\{A_{j,r}, B_{j,r}\} \leq c_6 n^{-\alpha/(2\alpha+1)} R^{1/(2\alpha+1)} M^{(\alpha+1-1/p)/((2\alpha+1)(1-\theta))}.$$

Moreover, this sum is equal to 0 if $2^r > 2\pi n M^{-1/(1-\theta)}$. Thus,

$$\sum_{r\geq 1} \sum_{\substack{j\geq 0\\2^{j}\leq \pi n 2^{1-r} M^{-1/(1-\theta)}}} \min\{A_{j,r}, B_{j,r}\} \leq c_{7}(\log n) n^{-\alpha/(2\alpha+1)} R^{1/(2\alpha+1)} M^{(\alpha+1-1/p)/((2\alpha+1)(1-\theta))}.$$

Note finally that we may bound

$$\sum_{\substack{j\geq 0}}\sum_{\substack{r\geq \ell+j+1\\ \frac{2r\leq \pi n}{2r\leq \pi n}}\mathbb{E}\left[\inf_{K_{j,r,\ell}\subset \widehat{\mathbb{Z}}_{j,r}(\ell)\cap \overline{\mathbb{Z}}_{j}}\widehat{T}_{j,r,\ell}(K_{j,r,\ell})\right]$$

from above as we did at the end of Section 4.5.

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A. A PROBABILITY INEQUALITY.

The following proposition is based on standard results in empirical processes. It uses the notion of VC subgraph classes, see [vdV13] for their definitions and properties.

Proposition 6. Let \mathcal{F} be an at most countable VC subgraph class of functions f defined on \mathbb{R} and with values in [-1,1]. We suppose that the VC dimension of \mathcal{F} is not larger than $d \ge 1$. We consider \overline{f} such that $|f(x)| \le \overline{f}(x) \le 1$ for all $x \in \mathbb{R}$. We further consider a map $\sigma^2(\cdot)$ defined on $\mathcal{F} \cup \{\overline{f}\}$ and satisfying $\sigma^2(f) \ge \mathbb{E}[f^2(X)]$ for all $f \in \mathcal{F} \cup \{\overline{f}\}$.

Then, there exists for all $\xi > 0$ an event of probability $1 - e^{-\xi}$ on which: for all $f \in \mathcal{F}$,

(67)
$$\frac{1}{n} \left| \sum_{i=1}^{n} \left(f(X_i) - \mathbb{E}[f(X_i)] \right) \right| \le C \min\{R_1(f), R_2(f)\},$$

where

(68)
$$R_{1}(f) = \sqrt{\frac{d}{n}\sigma^{2}(f)\log_{+}\left(\sigma^{2}(\bar{f})/\sigma^{2}(f)\right) + \frac{d}{n}\log_{+}\left(n\sigma^{2}(\bar{f})/d\right) + \sqrt{\sigma^{2}(f)\frac{\xi + \log n}{n}} + \frac{\xi + \log n}{n},$$
(69)
$$R_{2}(f) = \sqrt{\frac{1}{n}\sigma^{2}(f)\log_{-}\left(|\mathcal{F}|\right)} + \frac{1}{n}\log_{-}\left(|\mathcal{F}|\right) + \sqrt{\sigma^{2}(f)\frac{\xi + \log n}{n}} + \frac{\xi + \log n}{n}$$

(69)
$$R_2(f) = \sqrt{\frac{1}{n}}\sigma^2(f)\log_+(|\mathcal{F}|) + \frac{1}{n}\log_+(|\mathcal{F}|) + \sqrt{\sigma^2(f)\frac{\zeta + \log n}{n}} + \frac{\zeta + \log n}{n},$$

and where C is a numerical value. In the above inequality, $R_2(f) = +\infty$ if \mathcal{F} is infinite.

and where C is a numerical value. In the above inequality, $R_2(f) = +\infty$ if \mathcal{F} is infinite.

Proof of Proposition 6. We need the two following lemmas (see Theorem 3.1 of [GK06] for the first, and [Mas07] for the second).

Lemma 28. Let \mathcal{F} be an at most countable VC subgraph class of functions f defined on \mathbb{R} and with values in [-1,1]. We suppose that the VC dimension of \mathcal{F} is not larger than $d \geq 1$. We consider a map \bar{f} such that $|f(x)| \leq \bar{f}(x) \leq 1$ for all $x \in \mathbb{R}$. Let $\sigma^2(\bar{f})$ be a real number such that $\sigma^2(\bar{f}) \geq \mathbb{E}[\bar{f}^2(X)]$. Let moreover $\sigma^2 > 0$ be such that $\sup_{f \in \mathcal{F}} \mathbb{E}[f^2(X)] \leq \sigma^2$. Then,

$$\mathbb{E}\left[\sup_{f\in\mathcal{F}}\frac{1}{n}\left|\sum_{i=1}^{n}(f(X_i) - \mathbb{E}[f(X_i)])\right|\right] \le C\left[\sqrt{\frac{d}{n}\sigma^2\log_+(\sigma^2(\bar{f})/\sigma^2)} + \frac{d}{n}\log_+(\sigma^2(\bar{f})/\sigma^2)\right],$$

where C is a numerical value.

Lemma 29. Let \mathcal{F} be a finite class of functions f defined on \mathbb{R} and with values in [-1,1]. Let $\sigma^2 > 0$ such that $\sup_{f \in \mathcal{F}} \mathbb{E}[f^2(X)] \leq \sigma^2$. Then,

$$\mathbb{E}\left[\sup_{f\in\mathcal{F}}\frac{1}{n}\left|\sum_{i=1}^{n}(f(X_i) - \mathbb{E}[f(X_i)])\right|\right] \le C\left[\sqrt{\frac{1}{n}\sigma^2\log_+(|\mathcal{F}|)} + \frac{1}{n}\log_+(|\mathcal{F}|)\right],$$

where C is a numerical value.

We only prove the inequality with $R_1(f)$, the proof with $R_2(f)$ is similar (use Lemma 29 instead of Lemma 28 and set j_0 below as the smallest integer such that $2^{j_0-1} \ge n$). We may assume without loss of generality that $d \le n$ and $\sigma^2(f) \in (0,1]$ for all $f \in \mathcal{F}$. Let j_0 be the smallest integer such that $2^{j_0-1} \ge n/d$. We define for $j \in [1, j_0 - 1]$,

$$\mathcal{F}_j = \left\{ f \in \mathcal{F}, \, 2^{-j} < \sigma^2(f) \le 2^{-j+1} \right\},$$

and set

$$\mathcal{F}_{j_0} = \left\{ f \in \mathcal{F}, \, \sigma^2(f) \, \le 2^{-j_0+1} \right\}.$$

Let then for all $j \in [1, j_0]$,

$$\sigma_j^2 = 2^{-j+1}$$
$$Z_j = \sup_{f \in \mathcal{F}_j} \left| \frac{1}{n} \sum_{i=1}^n (f(X_i) - \mathbb{E}[f(X_i)]) \right|.$$

We consider some $\xi_0 > 0$ and deduce from Talagrand's inequality (see the second equation on page 170 of [Mas07]) that on an event $\Omega_j(\xi_0)$ of probability $1 - e^{-\xi_0}$,

$$Z_j \le 2\mathbb{E}[Z_j] + 2\sqrt{\left[2\sigma_j^2/n + 16\mathbb{E}[Z_j]/n\right]\xi_0} + 2\xi_0/n$$

Therefore, by using the elementary inequalities $\sqrt{x+y} \leq \sqrt{x} + \sqrt{y}$ and $2\sqrt{xy} \leq x+y$,

$$Z_j \le 6\mathbb{E}[Z_j] + 2\sqrt{2(\sigma_j^2/n)\xi_0} + 6\xi_0/n.$$

We now set for all $\sigma^2 > 0$,

$$A(\sigma^2) = \sqrt{\frac{d}{n}\sigma^2 \log_+(\sigma^2(\bar{f})/\sigma^2)} + \frac{d}{n}\log_+(\sigma^2(\bar{f})/\sigma^2)$$

Any $\sigma^2(f)$ belongs to $[(1/2)\sigma_j^2, \sigma_j^2]$ when $f \in \mathcal{F}_j$ with $j \leq j_0 - 1$. We deduce from Lemma 28 that for all such j,

$$\mathbb{E}[Z_j] \le CA(\sigma_j^2)$$

$$\le C\left[\sqrt{2}\sqrt{\frac{d}{n}\sigma^2(f)\log_+(\sigma^2(\bar{f})/\sigma^2(f))} + \frac{d}{n}\log_+\left(2n\sigma^2(\bar{f})/d\right)\right].$$

We deduce that on $\Omega_j(\xi_0)$: for all $f \in \mathcal{F}_j$,

$$\left| \frac{1}{n} \sum_{i=1}^{n} \left(f(X_i) - \mathbb{E}[f(X_i)] \right) \right| \le C' \left[\sqrt{\frac{d}{n} \sigma^2(f) \log_+(\sigma^2(\bar{f})/\sigma^2(f))} + \frac{d}{n} \log_+\left(n\sigma^2(\bar{f})/d\right) + \sqrt{(\sigma^2(f)/n)\xi_0} + \xi_0/n \right],$$

where C' is universal. When $j = j_0$, we rather have

$$\mathbb{E}[Z_j] \le CA(d/n)$$

$$\le C\left[\frac{d}{n}\sqrt{\log_+(n\sigma^2(\bar{f})/d)} + \frac{d}{n}\log_+\left(n\sigma^2(\bar{f})/d\right)\right]$$

$$\le 2C\frac{d}{n}\log_+\left(n\sigma^2(\bar{f})/d\right).$$

Hence, on $\Omega_{j_0}(\xi_0)$: for all $f \in \mathcal{F}_{j_0}$,

$$\left|\frac{1}{n}\sum_{i=1}^{n}\left(f(X_{i})-\mathbb{E}[f(X_{i})]\right)\right| \leq C''\left[\frac{d}{n}\log_{+}\left(n\sigma^{2}(\bar{f})/d\right)+\sqrt{(d/n^{2})\xi_{0}}+\xi_{0}/n\right]$$
$$\leq C'''\left[\frac{d}{n}\log_{+}\left(n\sigma^{2}(\bar{f})/d\right)+\xi_{0}/n\right],$$

where C''' is universal. We now set $\xi_0 = \xi + \log j_0 < \xi + 1.4 + \log n$ and conclude using a union bound.

B. AN ELEMENTARY LEMMA

Lemma 30. For all $a_1, a_2, b > 0, k \in [0, 1]$, and $r_0 \ge 1$,

$$\sum_{r=1}^{r_0} 2^{ra_1} \log_+^k \left(b 2^{-ra_2} \right) \le c 2^{r_0 a_1} \log_+^k \left(b 2^{-r_0 a_2} \right),$$

and

$$\sum_{r=r_0}^{\infty} 2^{-ra_1} \log_+^k (b2^{ra_2}) \le c2^{-r_0a_1} \log_+^k (b2^{r_0a_2}),$$

where c only depends on a_1, a_2, k .

Proof of Lemma 30. We only show the first inequality. The proof of the second inequality follows the same line. When k = 1, we write

$$\sum_{r=1}^{r_0} 2^{ra_1} \log_+ \left(b2^{-ra_2} \right) \le c_1 \left[\sum_{\substack{1 \le r \le r_0 \\ 2^{(r+1)a_2 \ge be^{-a_2/(a_1+1)}}} 2^{ra_1} + \sum_{\substack{1 \le r \le r_0 \\ 2^{(r+1)a_2 \le be^{-a_2/(a_1+1)}}} 2^{-ra_1} \log \left(b2^{-ra_2} \right) \right]$$

$$(70) \le c_2 \left[2^{r_0a_1} + \sum_{\substack{1 \le r \le r_0 \\ 2^{(r+1)a_2 \le be^{-a_2/(a_1+1)}}} 2^{-r} f(2^{-r}) \right],$$

where f denotes the map defined for x > 0 by $f(x) = x^{-a_1-1} \log(bx^{a_2})$. Let r_1 be the largest integer such that $2^{(r_1+1)a_2} < be^{-a_2/(a_1+1)}$. The proof when k = 1 is complete if $r_1 \leq 0$ and we assume from now on that $r_1 \geq 1$. We set $r_2 = \min\{r_0, r_1\}$. Since f is decreasing and non negative when $bx^{a_2} \geq e^{a_2/(a_1+1)}$, we get

$$\sum_{\substack{1 \le r \le r_0\\2^{(r+1)a_2} < be^{-a_2/(a_1+1)}}} 2^{-r} f(2^{-r}) \le 2 \sum_{r=1}^{r_2} \int_{2^{-r-1}}^{2^{-r}} f(x) \, \mathrm{d}x$$
$$\le 2 \int_{2^{-r_2-1}}^{\infty} f(x) \, \mathrm{d}x$$
$$\le c_3 2^{r_2a_1} \log_+(b2^{-a_2r_2})$$
$$\le c_4 2^{r_0a_1} \log_+(b2^{-a_2r_0}).$$

By putting this result in (70), we get the lemma when k = 1. The proof when $k \neq 1$ then follows from Hölder inequality:

$$\sum_{r=1}^{r_0} 2^{ra_1} \log_+^k \left(b2^{-ra_2} \right) = \sum_{r=1}^{r_0} 2^{rka_1} \log_+^k \left(b2^{-ra_2} \right) 2^{r(1-k)a_1}$$

$$\leq \left[\sum_{r=1}^{r_0} 2^{ra_1} \log_+ \left(b2^{-ra_2} \right) \right]^k \left[\sum_{r=1}^{r_0} 2^{ra_1} \right]^{1-k}$$

$$\leq c_5 2^{r_0ka_1} \log_+^k \left(b2^{-r_0a_2} \right) \times 2^{r_0(1-k)a_1}$$

$$\leq c_5 2^{r_0a_1} \log_+^k \left(b2^{-r_0a_2} \right).$$

C. Proof of Lemma 25

Proof of Lemma 25. Suppose that $p < \infty$ and consider an arbitrary real number q in $(1, p(\alpha + 1))$. We use (35) to get

$$\|\beta_{j,\cdot}\|_q^q \leq \frac{q}{q-p} \|\beta_{j,\cdot}\|_{\infty}^{q-p} \cdot \|\beta_{j,\cdot}\|_{p,\infty}^p.$$

Moreover, as ψ is bounded, $\|\beta_{j,\cdot}\|_{\infty} \leq c_1 2^{j/2}$. Therefore, using $f \in \mathcal{WB}_{p,\infty}^{\alpha}(R)$, we obtain for all $j \geq 0$

$$\|\beta_{j,\cdot}\|_q^q \le c_2 2^{j(1-p(\alpha+1)+q/2)},$$

where c_2 only depends on ψ, q, R . We deduce,

(71)
$$\sum_{j=0}^{\infty} 2^{j(q/2-1)} \sum_{k \in \mathbb{Z}} |\beta_{j,k}|^q < \infty.$$

Suppose that $p = \infty$ and take q > 1. Then, using that $|\beta_{j,k}| \le c_3 2^{j/2} f_{j,k}$, $\sum_{k \in \mathbb{Z}} f_{j,k} \le c_4$ (Lemma 4), and $\|\beta_{j,\cdot}\|_{\infty} \le c_5 2^{-j(\alpha+1/2)}$,

$$\|\beta_{j,\cdot}\|_q^q \leq c_6 2^{j(1+\alpha-(1/2+\alpha)q)},$$

and the sum (71) is finite.

Note that $|\alpha_k| \leq 1$ and $\sum_{k \in \mathbb{Z}} |\alpha_k| \leq 1$ and hence $\|\beta_{-1,\cdot}\|_q^q \leq 1$. Now, for all $j \geq 0$,

$$\begin{split} \left\| \sum_{k \in \mathbb{Z}} \beta_{j,k} \overline{\psi}_{j,k} \right\|_{q}^{q} &= 2^{j(q/2-1)} \int \left\| \sum_{k \in \mathbb{Z}} \beta_{j,k} \overline{\psi}(t-k) \right\|^{q} \mathrm{d}t \\ &\leq 2^{j(q/2-1)} \int \left\| \sum_{k \in \mathbb{Z}} |\beta_{j,k}| |\overline{\psi}(t-k)|^{1/q} |\overline{\psi}(t-k)|^{1-1/q} \right\|^{q} \mathrm{d}t \\ &\leq 2^{j(q/2-1)} \int \left(\sum_{k \in \mathbb{Z}} |\beta_{j,k}|^{q} |\overline{\psi}(t-k)| \right) \left(\sum_{k \in \mathbb{Z}} |\overline{\psi}(t-k)| \right)^{q-1} \mathrm{d}t \\ &\leq 2^{j(q/2-1)} \int \left(\sum_{k \in \mathbb{Z}} |\beta_{j,k}|^{q} |\overline{\psi}(t-k)| \right) \left\| \sum_{k \in \mathbb{Z}} |\overline{\psi}(t-k)| \right\|_{\infty}^{q-1} \mathrm{d}t \end{split}$$

$$\leq C2^{j(q/2-1)} \sum_{k \in \mathbb{Z}} |\beta_{j,k}|^q.$$

The same result holds true when j = -1. Let

$$\pi_J(f) = \sum_{j=-1}^J \sum_{k \in \mathbb{Z}} \beta_{j,k} \overline{\psi}_{j,k}.$$

The above ensures that $(\pi_J(f))_{J\geq 1}$ is a Cauchy sequence in $(\mathbb{L}^q, \|\cdot\|_q)$ and converges therefore to a map that must be f. In particular,

$$\|f\|_q \le \sum_{j=-1}^{\infty} \left\| \sum_{k \in \mathbb{Z}} \beta_{j,k} \overline{\psi}_{j,k} \right\|_q$$

,

hence the result.

D. Proof of Lemma 17

Proof of Lemma 17. For all u > 0,

(72)
$$\mathbb{P}\left[\min_{1 \le i \le n-1} (X_{(i+1)} - X_{(i)}) \le u\right] \le n \sup_{1 \le i \le n-1} \mathbb{P}\left[X_{(i+1)} - X_{(i)} \le u\right].$$

The density of $X_{(i+1)} - X_{(i)}$ is given for $x \ge 0$ by

$$\varphi_i(x) = \frac{n!}{(i-1)!(n-i-1)!} \int_{\mathbb{R}} F(t)^{i-1} (1 - F(t+x))^{n-i-1} f(t) f(t+x) \, \mathrm{d}t,$$

where F denotes the cumulative distribution function of X. We have for all u > 0,

$$\mathbb{P}\left[X_{(i+1)} - X_{(i)} \le u\right] = \frac{n!}{(i-1)!(n-i-1)!} \int_{\mathbb{R}} F(t)^{i-1} f(t) \left[\int_{0}^{u} (1 - F(t+x))^{n-i-1} f(t+x) \, \mathrm{d}x\right] \, \mathrm{d}t$$

$$= \frac{n!}{(i-1)!(n-i-1)!} \mathbb{E}\left[F(X_{1})^{i-1} (1 - F(X_{2}))^{n-i-1} 1_{X_{1} \le X_{2} \le X_{1}+u}\right]$$

$$\le \frac{n!}{(i-1)!(n-i-1)!} \mathbb{E}\left[F(X_{2})^{i-1} (1 - F(X_{2}))^{n-i-1} 1_{X_{1} \le X_{2} \le X_{1}+u}\right]$$

$$\le \frac{n!}{(i-1)!(n-i-1)!} \mathbb{E}\left[F(X_{2})^{i-1} (1 - F(X_{2}))^{n-i-1} \int_{X_{2}-u}^{X_{2}} f(t) \, \mathrm{d}t\right].$$

By using Hölder inequality,

$$\int_{X_2-u}^{X_2} f(t) \, \mathrm{d}t \le \|f\|_q u^{1-1/q}.$$

Moreover, $F(X_2)$ obeys to a uniform distribution on [0,1] as F is continuous and hence

$$\mathbb{E}\left[F(X_2)^{i-1}(1-F(X_2))^{n-i-1}\right] = \frac{(i-1)!(n-i-1)!}{n!}.$$

We use (72) to get

$$\mathbb{P}\left[\min_{1 \le i \le n-1} (X_{(i+1)} - X_{(i)}) \le u\right] \le n \|f\|_q u^{1-1/q}.$$

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We conclude by setting u such that $n \|f\|_q u^{1-1/q} = \xi/n$ and by using

$$2^{\hat{J}} \le \max\left\{1, 4L_{\psi} \min_{1 \le i \le n-1} \left(X_{(i+1)} - X_{(i)}\right)^{-1}\right\}.$$

E. PROOF OF PROPOSITION 3.

We consider some $\bar{k} \ge 1$ and endow $\{0,1\}^{2\bar{k}+1}$ with the Hamming distance Δ defined for $\delta, \delta' \in \{0,1\}^{2\bar{k}+1}$ by

$$\Delta(\delta, \delta') = \sum_{k=-\bar{k}}^{\bar{k}} |\delta_k - \delta'_k|.$$

Our proof relies on the classical Varshamov-Gilbert bound and Theorem 2.5 of [Tsy08].

Lemma 31. For all integer $\bar{k} \geq 8$, there exists a subset \mathscr{D} of $\{0,1\}^{2\bar{k}+1}$ such that

(73)
$$\log |\mathcal{D}| \ge 0.08k,$$

and such that

$$\sum_{k=-\bar{k}}^{\bar{k}} \delta_k = \bar{k}$$

for all $\delta \in \mathscr{D}$. Moreover, for all $\delta' \in \mathscr{D} \setminus \{\delta\}, \ \Delta(\delta, \delta') \ge \bar{k}/4.$

Sketch of the proof. We use Lemma 2.9 of [Tsy08] with $m = \bar{k}$. This leads to a subset Ω of $\{0, 1\}^{\bar{k}}$ and we set

$$\mathcal{D} = \left\{ (\delta_1, \dots, \delta_{2\bar{k}}, 0), (\delta_1, \dots, \delta_{\bar{k}}) \in \Omega, \forall k \in \{1, \dots, \bar{k}\}, \delta_{\bar{k}+k} = 1 - \delta_k \right\}.$$

Lemma 32. Let \mathscr{D} be a subset of $\{0,1\}^{2\bar{k}+1}$ containing at least 3 elements. We consider $\eta > 0$ and a family of densities $\mathscr{F} = \{f_{\delta}, \delta \in \mathscr{D}\}$ satisfying

(74)
$$d_1(f_{\delta}, f_{\delta'}) \ge \eta$$

for all $\delta \neq \delta' \in \mathscr{D}$. Moreover, we suppose

(75)
$$K(f_{\delta}, f_{\delta'}) \le \frac{\log |\mathcal{D}|}{16n}$$

for all $\delta, \delta' \in \mathscr{D}$ such that $\delta \neq \delta'$, where $K(\cdot, \cdot)$ denotes the Kullback Leibler divergence defined by

$$K(f_{\delta}, f_{\delta'}) = \int f_{\delta}(x) \log \left(f_{\delta}(x) / f_{\delta'}(x) \right) \, \mathrm{d}x.$$

Then,

(76)
$$\inf_{\tilde{f}} \sup_{f \in \mathscr{F}} \mathbb{E}\left[d_1(f, \tilde{f})\right] \ge 0.13\eta,$$

where the infimum is taken over all estimators \tilde{f} .

We consider $\bar{k} \ge 14$ and the set \mathscr{D} given by Lemma 31. The whole point of the proof is to build a family $\mathscr{F} = \{f_{\delta}, \delta \in \mathscr{D}\}$ of densities satisfying the requirements of the preceding lemma. Note that we already have $|\mathscr{D}| \ge 3$.

Let $\ell \geq 1$ be the smallest integer such that $\operatorname{supp} \bar{\psi}$ and $\operatorname{supp} \bar{\phi}$ are both included in $(-2^{\ell}, 2^{\ell})$. We consider integers $j \geq -1$, $j_0 \geq 0$ such that $2^{j_0+j-\ell} \geq 29$. We define \bar{k} as the largest integer satisfying $\bar{k} \leq (2^{j_0+j-\ell}-1)/2$. Note that $\bar{k} \geq 14$ since $2^{j_0+j-\ell} \geq 29$.

We consider b > 0 and introduce the map h_{δ} for $\delta \in \mathscr{D}$ and $x \in \mathbb{R}$ by

$$h_{\delta}(x) = b \sum_{k=-\bar{k}}^{k} \delta_k \bar{\psi}_{j,2^{\ell+1}k}(x).$$

There exists a compactly supported density g_0 on [-2, 2] satisfying $\inf_{x \in [-1,1]} g_0(x) \ge 1/4$, $||g_0||_{\infty} \le 1$, and belonging to $\mathcal{B}^{\alpha}_{p,\infty}(R_{g_0})$ for some $R_{g_0} > 0$. We then set for $x \in \mathbb{R}$,

$$g(x) = 15^{-1/\theta} M^{1/\theta} 2^{-j_0/\theta} g_0(2^{-j_0}x)$$

When R is large enough, we may consider a density $\zeta \in \mathcal{B}_{p,\infty}^{\alpha}(R/\max\{3^{1/p},3\})$, bounded by 1, vanishing on $[-2^{j_0+1}, 2^{j_0+1}]$, and compactly supported on an interval of length 2. We have $\zeta \in \mathcal{T}_{\theta}(M/3)$ for M large enough. We set

$$q = 15^{-1/\theta} M^{1/\theta} 2^{j_0(1-1/\theta)} + b\bar{k} 1_{i=-1},$$

and define for $\delta \in \mathscr{D}$ and $x \in \mathbb{R}$,

$$f_{\delta}(x) = (1-q)\zeta(x) + g(x) + h_{\delta}(x).$$

We show after the present proof:

Lemma 33. Suppose that the parameters $j \ge -1$, $j_0 \ge 0$, b > 0 are chosen in such a way that $2^{j_0+j-\ell} \ge 29$, and such that

(77)
$$b2^{j_0/p}2^{j(\alpha+1/2)}1_{j\geq 0} \le a_1R$$

(78)
$$M^{1/\theta} 2^{j_0(1/p-1/\theta-\alpha)} \le a_2 R$$

(79)
$$b2^{j/2}2^{j_0/\theta} \le a_3 M^{1/\theta}$$

$$(80) 2^{j_0/\theta} b^2 < a_4 M^{1/\theta} n^{-1}$$

$$(81) q \le 1.$$

In the above conditions, a_1, a_2, a_3, a_4 are suitable terms depending only on p, θ, g_0 and the wavelet basis.

Then, $\mathscr{F} = \{f_{\delta}, \delta \in \mathscr{D}\}\$ is a collection of densities included in $\mathcal{B}^{\alpha}_{p,\infty}(R) \cap \mathcal{T}_{\theta}(M)$. It satisfies (75) and (74) with $\eta = cb2^{j_0+j/2}$ and c depending on the wavelet basis only. In particular,

(82)
$$\inf_{\tilde{f}} \sup_{\mathcal{B}_{p,\infty}^{\alpha}(R) \cap \mathcal{T}_{\theta}(M)} \mathbb{E}\left[d_{1}(f,\tilde{f})\right] \geq 0.13cb2^{j_{0}+j/2}$$

We first prove when $\theta < p$

(83)
$$\inf_{\tilde{f}} \sup_{f \in \mathcal{B}_{p,\infty}^{\alpha}(R) \cap \mathcal{T}_{\theta}(M)} \mathbb{E}\left[d_{1}(f,\tilde{f})\right] \geq cR^{(1-\theta)/(1+\alpha-\theta/p)} M^{(\alpha+1-1/p)/(\alpha+1-\theta/p)} n^{-\alpha(1-\theta)/(\alpha+1-\theta/p)}.$$

For this, we define j_0 as the smallest integer such that

$$2^{j_0\left[\frac{1}{p}-\frac{\alpha+1}{\theta}\right]} \leq \frac{R}{M^{\frac{\alpha+1}{\theta}}n^{\alpha}}.$$

Note that j_0 tends to infinity when n grows up. Therefore the condition $2^{j_0+j-\ell} \ge 29$ is satisfied when n is large enough. We define c_1 and c_2 small enough so that $c_1^{\alpha+1}c_2 \le a_1, c_1c_2^2 \le a_4, c_1c_2 \le a_3$. Since $\theta < p$, we may take n large enough and define the largest integer $j \ge 0$ such that

$$2^{j} \leq c_1 M^{1/\theta} n 2^{-j_0/\theta}$$

We define

$$b = c_2 2^{j/2} n^{-1}.$$

We may check that (77), (78), (79) and (80) are fulfilled. Note that q tends to 0 when n grows up and is therefore smaller than 1 when n is large enough. We then deduce (83) from (82).

When $\theta = p$, the above reasoning works when $R^p \ge M/c_1^{\alpha p}$. This condition ensures the existence of $j \ge 0$. It is worth mentioning that the lower bound

$$R^{(1-\theta)/(1+\alpha-\theta/p)}M^{(\alpha+1-1/p)/(\alpha+1-\theta/p)}n^{-\alpha(1-\theta)/(\alpha+1-\theta/p)}$$

follows from

(84)
$$\inf_{\tilde{f}} \sup_{f \in \mathcal{B}^{\alpha}_{p,\infty}(R) \cap \mathcal{T}_{\theta}(M)} \mathbb{E}\left[d_1(f,\tilde{f})\right] \ge cM \mathbf{1}_{\gamma=1-\theta} n^{-\gamma}.$$

when $\theta = p$ and $R^p < M/c_1^{\alpha p}$. The proof of which is given below.

We now show (84). We set j = -1, consider c_1, c_2 such that $c_2 \leq a_4$ and $\sqrt{c_1 c_2/2} \leq a_3$, define j_0 as the largest integer such that

$$2^{j_0/\theta} \leq c_1 M^{1/\theta} n,$$

and b^2 such that

$$2^{j_0/\theta}b^2 = c_2 M^{1/\theta} n^{-1}.$$

We may check that the conditions of the lemma are met hence the result.

We finally prove

$$\inf_{\tilde{f}} \sup_{f \in \mathcal{B}^{\alpha}_{p,\infty}(R) \cap \mathcal{T}_{\theta}(M)} \mathbb{E} \left[d_1(f,\tilde{f}) \right] \\ \geq c R^{1/(2\alpha+1)} M^{(\alpha+1-1/p)/((1-\theta)(2\alpha+1))} n^{-\alpha/(2\alpha+1)}.$$

We define $j_0 \ge 0$ such that $2^{j_0-1} \le c_1^{\theta/(1-\theta)} M^{1/(1-\theta)} \le 2^{j_0}$ where $c_1 = 2 \cdot 15^{-1/\theta}$. We define j as the smallest integer such that

$$M^{-1/\theta}R^2 2^{-j(2\alpha+1)} < n^{-1} 2^{j_0(2/p-1/\theta)}.$$

Since j tends to $+\infty$ when n grows up, the condition $2^{j_0+j-\ell} \ge 29$ is fulfilled when n is large enough. Besides, $q \le 1/2$ and (78) holds true if R_0, M_0 are large enough. Moreover, we set

$$b = c_2 R 2^{-j_0/p} 2^{-j(\alpha+1/2)}$$

where $c_2 = \min \{a_1, \sqrt{a_4}\}$. We conclude by applying Lemma 33 as above.

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E.1. Proof of Lemma 33.

Lemma 34. For all $\delta \in \mathscr{D}$, the map h_{δ} is compactly supported on $[-2^{j_0}, 2^{j_0}]$ and belongs to

$$\mathcal{B}_{p,\infty}^{\alpha}\left(b\bar{k}^{1/p}2^{j(\alpha+1/2-1/p)}1_{j\geq 0}\right).$$

Moreover, $|h_{\delta}| \in \mathcal{T}_{\theta}(c_1 b^{\theta} 2^{j\theta/2} 2^{j_0})$, where c_1 depends on θ and the wavelet basis only,

(85)
$$\int h_{\delta}(x) \,\mathrm{d}x = b\bar{k}\mathbf{1}_{j=-1}$$

and

 $\|h_{\delta}\|_{\infty} \le c_2 b 2^{j/2},$

where c_2 only depends on the wavelet basis.

Sketch of proof of Lemma 34. We remark that the supports of $\bar{\psi}_{j,2^{\ell+1}k}$ and $\bar{\psi}_{j,2^{\ell+1}k'}$ are disjoint when $k \neq k'$. Therefore,

$$\|h_{\delta}\|_{\infty} \le b \sup_{k \in \mathbb{Z}} \|\bar{\psi}_{j,2^{\ell+1}k}\|_{\infty} \le b \max\{\|\bar{\phi}\|_{\infty}, 2^{j/2} \|\bar{\psi}\|_{\infty}\}.$$

We then deduce from Lemma 2.1 of [CL20],

$$|h_{\delta}| \in \mathcal{T}_{\theta}\left(\|h_{\delta}\|_{\infty}^{\theta}(2^{j_0+1}+1)\right)$$

which shows $|h_{\delta}| \in \mathcal{T}_{\theta} \left(c_1 b^{\theta} 2^{(j/2)\theta} 2^{j_0} \right)$.

We get (85) by noticing that $\int \bar{\phi} = 1$ and $\int \bar{\psi} = 0$ as the wavelet basis is bi-orthogonal and that $\phi = 1_{[0,1]}$.

Lemma 35. The map g is compactly supported on $[-2^{j_0+1}, 2^{j_0+1}]$. It satisfies

$$\int g(x) \, \mathrm{d}x = 15^{-1/\theta} M^{1/\theta} 2^{j_0(1-1/\theta)}.$$

Besides, for all $x \in [-2^{j_0}, 2^{j_0}]$

$$g(x) \ge \left(4 \cdot 15^{1/\theta}\right)^{-1} M^{1/\theta} 2^{-j_0/\theta}.$$

It belongs to

$$\mathcal{B}_{p,\infty}^{\alpha}\left(c_{3}M^{1/\theta}2^{j_{0}(1/p-1/\theta-\alpha)}\right)\bigcap\mathcal{T}_{\theta}(M/3),$$

where c_3 only depends on g_0, θ .

Sketch of the proof of Lemma 35. We only prove that $g \in \mathcal{B}_{p,\infty}^{\alpha}(c_3 M^{1/\theta} 2^{j_0(1/p-1/\theta-\alpha)})$. The wavelet coefficient of g is denoted for $j' \geq 0$ and $k' \in \mathbb{Z}$ by

$$\beta'_{j',k'} = \int g(x)\psi_{j',k'}(x) \,\mathrm{d}x.$$

Yet,

$$\beta_{j',k'}' = 15^{-1/\theta} M^{1/\theta} 2^{(1/2 - 1/\theta)j_0} \int g_0(x) \psi_{j'+j_0,k'}(x) \, \mathrm{d}x$$

In particular,

$$\|\beta'_{j',\cdot}\|_p \le 15^{-1/\theta} M^{1/\theta} 2^{j_0(1/p-1/\theta-\alpha)} R_{g_0} 2^{-j'(\alpha+1/2-1/p)}$$

hence the result.

Proof of Lemma 33. By choosing a_1 and a_2 appropriately, we deduce from (77) and (78) that gand h_{δ} lie in $\mathcal{B}_{p,\infty}^{\alpha}(R/\max\{3^{1/p},3\})$. We deduce from (81) and from the (quasi) triangle inequality for the (quasi) norms $\|\cdot\|_p$ that $f_{\delta} \in \mathcal{B}_{p,\infty}^{\alpha}(R)$. Note that (79) yields $g(x) \geq 2\|h_{\delta}\|_{\infty}$ for all $x \in [-2^{j_0}, 2^{j_0}]$ if a_3 is suitably taken. In particular, f_{δ} is non-negative and is therefore a density. Moreover, (79) implies $|h_{\delta}| \in \mathcal{T}_{\theta}(M/3)$ and hence $f_{\delta} \in \mathcal{T}_{\theta}(M)$. It satisfies

$$f_{\delta}(x) \ge \left(8 \cdot 15^{1/\theta}\right)^{-1} M^{1/\theta} 2^{-j_0/\theta}$$

for all $x \in [-2^{j_0}, 2^{j_0}]$. Moreover, $f_{\delta}(x) = (1-q)\zeta(x) + g(x)$ for all $x \notin [-2^{j_0}, 2^{j_0}]$.

We recall the classical inequality

$$K(f_{\delta}, f_{\delta'}) \leq \int \frac{(f_{\delta}(x) - f_{\delta'}(x))^2}{\min\{f_{\delta}(x), f_{\delta'}(x)\}} \,\mathrm{d}x.$$

We deduce that for all $\delta \neq \delta' \in \mathscr{D}$ (note that the supports of $\bar{\psi}_{i,2^{\ell+1}k}$ are disjoint when k varies),

$$\begin{split} K(f_{\delta}, f_{\delta'}) &\leq 8 \cdot 15^{1/\theta} 2^{j_0/\theta} M^{-1/\theta} \times b^2 \sum_{k=-\bar{k}}^k |\delta_k - \delta'_k|^2 \times \max\left\{ \int \ \bar{\phi}^2(x) \, \mathrm{d}x, \int \ \bar{\psi}^2(x) \, \mathrm{d}x \right\} \\ &\leq 16 \cdot 15^{1/\theta} 2^{j_0/\theta} M^{-1/\theta} b^2 \bar{k} \max\left\{ \int \ \bar{\phi}^2(x) \, \mathrm{d}x, \int \ \bar{\psi}^2(x) \, \mathrm{d}x \right\}. \end{split}$$

We may therefore fulfil condition (75) by choosing a_4 thanks to (80) and (73). Moreover,

$$d_1(f_{\delta}, f_{\delta'}) = b \sum_{k=-\bar{k}}^{\bar{k}} |\delta_k - \delta_{k'}| \int \left| \bar{\psi}_{j, 2^{\ell+1}k}(x) \right| \, \mathrm{d}x$$

$$\geq b \; (\bar{k}/4) \min\left\{ 2^{-j/2} \int |\bar{\psi}(x)| \, \mathrm{d}x, \int |\bar{\phi}(x)| \, \mathrm{d}x \right\}.$$

We use the definition of k to conclude.

F. Proof of Proposition 4

We only need to prove the proposition when $\alpha = 1/p - 1$, what we assume below. We define $j_0 \geq 2$ as the smallest integer such that $(n^p + 1)2^{-j_0} \leq 1/4$ and consider $j_1 \geq j_0$.

We set for $j \in \{j_0, \dots, j_1\}, k_j = 2(n^p + 1)(2^{j-j_0} - 1)$ and

$$K_j = \{k_j, k_j + 1, k_j + 2, \dots, k_j + n^p\}.$$

We introduce for $k \in \mathbb{Z}$, $I_{j,k} = [k2^{-j}, (k+1)2^{-j}]$. Note that $I_{j,k} \subset [0, 1/2]$ for all j between j_0 and j_1 and all $k \in K_j$. Moreover, $I_{j,k} \cap I_{j',k'} = \emptyset$ if $j \neq j'$, no matter $(k,k') \in K_j \times K_{j'}$.

We set $D = (j_1 - j_0 + 1)(n^p + 1)$ and write the elements δ of $\{0, 1\}^D$ as $\delta = (\delta_{j,k})_{j \in \{j_0, \dots, j_1\}, k \in K_j}$. We endow $\{0, 1\}^D$ with the Hamming distance defined for all $\delta, \delta' \in \{0, 1\}^D$ by

$$\Delta(\delta, \delta') = \sum_{j=j_0}^{j_1} \sum_{k \in K_j} |\delta_{j,k} - \delta'_{j,k}|.$$

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We set for all $\delta \in \{0,1\}^D$ and $x \in \mathbb{R}$,

$$\varphi_{\delta}(x) = \frac{1}{D} \sum_{j=j_0}^{j_1} 2^j \sum_{k \in K_j} \delta_{j,k} \mathbf{1}_{I_{j,k}}(x).$$

We show below after the present proof:

Lemma 36. For all $\varepsilon > 0$, D large enough, and $\delta \in \{0,1\}^D$, φ_{δ} belongs to $\mathfrak{B}^{\alpha}_{p,\infty}(\varepsilon)$.

We define for $x \in \mathbb{R}$,

$$f_{\delta}(x) = \varphi_{\delta}(x) + \varphi_{1-\delta}(x - 1/2).$$

Thereby, f_{δ} is a compactly supported density on [0,1] lying in $\mathfrak{B}_{p,\infty}^{\alpha}(R)$ if D is large enough. Moreover, for all $\delta, \delta' \in \{0,1\}^D$,

$$d_1(f_{\delta}, f_{\delta'}) = \frac{2}{D} \Delta(\delta, \delta').$$

Recall that the square h^2 of the Hellinger distance is not larger than half of d_1 . We conclude by applying Assouad's lemma and by choosing j_1 large enough (see Lemma 2 of [Bir06b]).

Proof of Lemma 36. We have for all h > 0,

$$\int |\varphi_{\delta}(t+h) - \varphi_{\delta}(t)|^{p} dt \leq \frac{1}{D^{p}} \sum_{j=j_{0}}^{j_{1}} 2^{jp} \sum_{k \in K_{j}} \int |1_{I_{j,k}}(t+h) - 1_{I_{j,k}}(t)|^{p} dt$$
$$\leq \frac{2}{D^{p}} \sum_{j=j_{0}}^{j_{1}} 2^{jp} \sum_{k \in K_{j}} \min \left\{ 2^{-j}, h \right\}$$
$$\leq \frac{2(n^{p}+1)}{D^{p}} \sum_{j=j_{0}}^{j_{1}} 2^{jp} \min \left\{ 2^{-j}, h \right\}$$
$$\leq c \frac{n^{p}+1}{D^{p}} h^{1-p}$$
$$\leq c \frac{n^{p}+1}{D^{p}} h^{\alpha p}.$$

Consider now some odd number r larger than α and note that

$$2\Delta_h^r(\varphi_\delta)(t) = \sum_{k=0}^r \binom{r}{k} (-1)^k \left[\varphi_\delta(t+kh) - \varphi_\delta(t+(r-k)h)\right].$$

The above entails that for all D large enough,

$$\int |\Delta_h^r(\varphi_\delta)(t)|^p \, \mathrm{d}t \le (\varepsilon/2)^p h^{\alpha p}$$

By noticing that

$$\|\varphi_{\delta}\|_{p}^{p} \leq \frac{n^{p}+1}{D^{p}} \sum_{j=j_{0}}^{j_{1}} 2^{-j(1-p)} \leq (\varepsilon/2)^{p}$$

when D is large enough, we conclude that φ_{δ} belongs to $\mathfrak{B}_{p,\infty}^{\alpha}(\varepsilon)$.

G. PROOF OF PROPOSITION 1.

If f is a density, $F_{j,k} \leq 1$ and

$$\sum_{k\in\mathbb{Z}}F_{j,k}=1.$$

We deduce from the elementary inequality

$$\left(\sum_{k\in\mathbb{Z}}F_{j,k}^{\theta}\right)^{1/\theta}\geq\sum_{k\in\mathbb{Z}}F_{j,k}$$

that M must satisfy $M \ge 1$ if $f \in \mathcal{T}_{\theta}(M)$. This proves the first assertion.

We now show the first part of the second point. We deduce from (2) that for all $j \ge 0$ and $k \in \mathbb{Z}$,

$$F_{j,k} = \sum_{j'=-1}^{\infty} \sum_{k' \in \mathbb{Z}} \beta_{j',k'} \int_{2^{-j}(k-1/2)}^{2^{-j}(k+1/2)} \bar{\psi}_{j',k'}.$$

Let $\bar{L} > 0$ such that $[-\bar{L}, \bar{L}]$ contains the supports of $\bar{\phi}$ and $\bar{\psi}$. The above integral is zero if $k \notin K_{j',k'}$ where

$$K_{j',k'} = \left[-1/2 + 2^{j-\max\{j',0\}} \left(k' - \bar{L}\right), 1/2 + 2^{j-\max\{j',0\}} \left(k' + \bar{L}\right)\right].$$

We deduce

$$F_{j,k} \le c_1 \sum_{j'=-1}^{\infty} \sum_{k' \in \mathbb{Z}} \left| \beta_{j',k'} \right| \mathbf{1}_{k \in K_{j',k'}} 2^{-\max\{j-j'/2,j'/2\}},$$

,

where c_1 depends on $\bar{\phi}, \bar{\psi}$ only. Now, using the same elementary inequality as previously,

$$\begin{split} \sum_{k\in\mathbb{Z}} F_{j,k}^p &\leq c_1^p \sum_{k\in\mathbb{Z}} \sum_{j'=-1}^{\infty} \sum_{k'\in\mathbb{Z}} \left| \beta_{j',k'} \right|^p \mathbf{1}_{k\in K_{j',k'}} 2^{-p\max\{j-j'/2,j'/2\}} \\ &\leq c_1^p \sum_{j'=-1}^{\infty} \sum_{k'\in\mathbb{Z}} \left| K_{j',k'} \right| \left| \beta_{j',k'} \right|^p 2^{-p\max\{j-j'/2,j'/2\}}, \\ &\leq c_2 \sum_{j'=-1}^{\infty} \sum_{k'\in\mathbb{Z}} \left| \beta_{j',k'} \right|^p \left[2^{j(1-p)+j'(p/2-1)} + 2^{-j'p/2} \right] \\ &\leq c_2 R^p \sum_{j'=-1}^{\infty} \left[2^{j(1-p)} 2^{-j'p\alpha} + 2^{-j'p(\alpha+1-1/p)} \right] \\ &\leq c_3 R^p 2^{j(1-p)}, \end{split}$$

which gives the first part of the second point.

The proof of the second part of the second point, as well as the proof of the third point is quite easy, and we move directly to the proof of the fourth point. For all $k \notin [-2^j - 1/2, 2^j + 1/2]$,

$$F_{j,k} \leq A^b \int_{2^{-j}(k-1/2)}^{2^{-j}(k+1/2)} |x|^{-b} \, \mathrm{d}x.$$

In particular, for all $k \ge 2^j + 1/2$,

$$F_{j,k} \leq A^b \int_{2^{-j}(k-1/2)}^{2^{-j}(k+1/2)} \left[2^{-j}(k-1/2) \right]^{-b} \, \mathrm{d}x \leq A^b 2^{-j(1-b)} \left[(k-1/2) \right]^{-b}$$

and for all $k \le -2^j - 1/2$,

$$F_{j,k} \leq A^{b} 2^{-j(1-b)} \left[(-k - 1/2) \right]^{-b}.$$

The number of $k \notin [-2^j - 1/2, 2^j + 1/2]$ such that $F_{j,k} \ge t$ is therefore not larger than $2 \left[4t^{-1/b} - i(1/b-1) + 1 \right]$

$$2\left\lfloor At^{-1/b}2^{-j(1/b-1)}+1\right\rfloor.$$

Moreover, the number of $k \in [-2^j - 1/2, 2^j + 1/2]$ such that $F_{j,k} \ge t$ is not larger than

$$t^{-1/b} \sum_{k \in [-2^{j} - 1/2, 2^{j} + 1/2]} F_{j,k}^{1/b} \le t^{-1/b} \left(\sum_{k \in \mathbb{Z}} F_{j,k} \right)^{1/b} [2^{j+1} + 2]^{1-1/b} \le t^{-1/b} [2^{j+1} + 2]^{1-1/b}.$$

Now,

$$\sup_{t>0} \left\{ t^{1/b} \sum_{k\in\mathbb{Z}} 1_{F_{j,k}\geq t} \right\} = \sup_{t\in(0,1]} \left\{ t^{1/b} \sum_{k\in\mathbb{Z}} 1_{F_{j,k}\geq t} \right\}$$
$$\leq \left(2^{j+1}+2\right)^{1-1/b} + 2\left(A2^{-j(1/b-1)}+1\right),$$

which concludes the proof.

H. PROOF OF EQUALITY (1) IN $(\mathbb{L}^1(\mathbb{R}), d_1)$.

The following arguments come mainly from [HKPT12]. We introduce the Kernel K defined for $x, y \in \mathbb{R}$ by

$$K(x,y) = \sum_{k \in \mathbb{Z}} \bar{\phi}(x-k)\phi(y-k).$$

We put for $J \ge 0$, $K_J(x, y) = 2^J K(2^J x, 2^J y)$ and

$$K_J f(x) = \int_{\mathbb{R}} K_J(x, y) f(y) \, \mathrm{d}y$$

We also set

$$K'_J f = \sum_{k \in \mathbb{Z}} \alpha_k \bar{\phi}_k + \sum_{j=0}^J \sum_{k \in \mathbb{Z}} \beta_{j,k} \bar{\psi}_{j,k}.$$

When $f \in \mathbb{L}^2(\mathbb{R})$, $K_J f$ is the (oblic) projection of f on the space spanned by the basis $(\bar{\phi}_{J,k})_{k\in\mathbb{Z}}$ where $\bar{\phi}_{J,k}(x) = 2^{J/2}\bar{\phi}(2^J x - k)$. Therefore, $K_J f = K'_J f$ for all $f \in \mathbb{L}^2(\mathbb{R})$.

Since ϕ and $\overline{\phi}$ are two compactly supported bounded functions, there exists an integrable function F such that $|K(x,y)| \leq F(x-y)$ for all $x, y \in \mathbb{R}$. This entails that K_J is continuous in $(\mathbb{L}^1(\mathbb{R}), d_1)$. The same thing holds true for K'_J and hence, by using a density argument, $K_J f = K'_J f$ for all $f \in \mathbb{L}^1(\mathbb{R})$.

Since $K_J f \to f$ in $(\mathbb{L}^2(\mathbb{R}), d_2)$, we have for all $f \in \mathbb{L}^2(\mathbb{R})$

$$\int_{\mathbb{R}} \left(\int_{\mathbb{R}} K(x2^J, y) \, \mathrm{d}y - 1 \right)^2 f^2(x) \, \mathrm{d}x \to 0 \quad \text{when } J \to +\infty.$$

We deduce (see Lemma 8.4 of [HKPT12]) that $\int_{\mathbb{R}} K(x, y) \, dy = 1$ for all $x \in \mathbb{R}$.

Therefore, we have for all $f \in \mathbb{L}^1(\mathbb{R})$,

$$\begin{aligned} \|K_J f - f\|_1 &\leq \int_{\mathbb{R}} |K_J(x, y)| \, |f(y) - f(x)| \, \mathrm{d}x \, \mathrm{d}y \\ &\leq \int_{\mathbb{R}} F(t) \left(\int_{\mathbb{R}} \left| f(x + 2^{-J}t) - f(x) \right| \, \mathrm{d}x \right) \, \mathrm{d}t. \end{aligned}$$

$$f \text{ in } (\mathbb{L}^1(\mathbb{R}), d_1). \qquad \Box$$

This entails $K_J f \to f$ in $(\mathbb{L}^1(\mathbb{R}), d_1)$.

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