Moment bounds for dependent sequences in smooth Banach spaces

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Abstract

We prove a Marcinkiewicz-Zygmund type inequality for random variables taking values in a smooth Banach space. Next, we obtain some sharp concentration inequalities for the empirical measure of $\{T, T^2, \dots, T^n\}$, on a class of smooth functions, when T belongs to a class of nonuniformly expanding maps of the unit interval.

1 Introduction and notations

Let $(\mathbb{B}, |\cdot|_{\mathbb{B}})$ be a real separable Banach space. The notion of *p*-smooth Banach spaces (1 was introduced in a famous paper by Pisier ([17], Section 3). These spaces play the same role with respect to martingales as spaces of type*p*do with respect to the sums of independent random variables.

We shall follow the approach of Pinelis [16], who showed that 2-smoothness is in some sense equivalent to a control of the second directional derivative of the map ψ_2 defined by $\psi_2(x) = |x|_{\mathbb{B}}^2$. In particular, if there exists C > 0 such that, for any x, u in \mathbb{B} ,

$$D_{u,u}^2 \psi_2(x) \le C |u|_{\mathbb{B}}^2, \tag{1.1}$$

then the space \mathbb{B} is 2-smooth (here $D^2_{u,v}g(x)$ denotes the second derivative of g at point x, in the directions u, v). In his 1994 paper, Pinelis [16] used the property (1.1) to derive Burkholder and Rosenthal moment inequalities as well as exponential bounds for \mathbb{B} -valued martingales.

In this paper, we shall consider a class of smooth Banach spaces, whose smoothness property is described as follows. Let p be a real number in $[2, \infty[$ and let ψ_p be the function from \mathbb{B} to \mathbb{R} defined by

$$\psi_p(x) = |x|_{\mathbb{B}}^p \,. \tag{1.2}$$

We say that the real separable Banach space $(\mathbb{B}, |\cdot|_{\mathbb{B}})$ belongs to the class $\mathcal{C}_2(p, c_p)$ if the function ψ_p is two times Fréchet differentiable and satisfies, for all x and u in \mathbb{B} ,

$$D^{2}\psi_{p}(x)(u,u) \leq c_{p}|x|_{\mathbb{B}}^{p-2}|u|_{\mathbb{B}}^{2}.$$
(1.3)

Here $D^2g(x)$ denotes the usual second order Fréchet derivative of g at point x. Using the chain rule, it is easy to see that if (1.1) holds for the second order Fréchet derivative and if $D\psi_2(0) = 0$, then (1.3) is satisfied. Before describing our results, let us quote that the class $C_2(p, c_p)$ contains the \mathbb{L}^q -spaces for $q \geq 2$, for which one can compute the constant c_p . The following lemma will be proved in Appendix.

Lemma 1.1.

- 1. For any $q \in [2, \infty[$ and any measure space $(\mathcal{X}, \mathcal{A}, \mu)$, the space $\mathbb{L}^q = \mathbb{L}^q(\mathcal{X}, \mathcal{A}, \mu)$ belongs to the class $\mathcal{C}_2(p, c_p)$ with $c_p = p(\max(p, q) 1)$.
- 2. If \mathbb{B} is a separable Hilbert space then it belongs to the class $\mathcal{C}_2(p, c_p)$ with $c_p = p(p-1)$.

The main result of this paper is a Marcinkiewicz-Zygmund type inequality for the moment of order p of partial sums S_n of \mathbb{B} -valued random variables, when \mathbb{B} belongs to the class $C_2(p, c_p)$. The upper bound is expressed in terms of conditional expectations of the random variables with respect to a past σ -field, and extends the corresponding upper bound by Dedecker and Doukhan [3] for real-valued random variables. As in [18] and [3], the proof is done by writing $\psi_p(S_n)$ as a telescoping sum. The property (1.3) enables to use the Taylor integral formula at order 2 to control the terms of the telescoping sums.

This Marcinkiewicz-Zygmund type bound together with the Rosenthal type bound given in [6] and the deviation inequality given in [5] provide a full description of the moment bounds for sums of \mathbb{B} -valued random variables, when \mathbb{B} belongs to the class $\widetilde{C}_2(p, c_p)$. As we shall see, these bounds apply to a large class of dependent sequences, in the whole range from short to long dependence.

As an application, we shall focus on the \mathbb{L}^q -norm of the centered empirical distribution function G_n of the iterates of a nonuniformly expanding map T of the unit interval (modelled by a Young tower with polynomial tails). On the probability space [0,1] equipped with the T-invariant probability ν , the covariance between two Hölder observables of T and T^n is of order $n^{-(1-\gamma)/\gamma}$ for some $\gamma \in (0,1)$. Hence the sequence of the iterates $(T^i)_{i\geq 1}$ is short-range dependent if $\gamma < 1/2$ and long-range dependent if $\gamma \in [1/2, 1)$. The moment and deviation bounds for the \mathbb{L}^q -norm of G_n are given in Theorem 4.1 in the short range dependent case, and in Theorems 4.2 and 4.3 in the long range dependent case. In Remark 4.1, we give some arguments, based on a limit theorem for the \mathbb{L}^2 -norm of G_n , showing that the deviation bounds of Theorem 4.3 are in some sense optimal.

As a consequence of these results, we obtain in Corollary 4.1 a full description of the behavior of $||W_1(\nu_n,\nu)||_p$ for $p \ge 1$, where $W_1(\nu_n,\nu)$ is the Wasserstein distance between the empirical measure ν_n of $\{T, T^2, \ldots, T^n\}$ and the invariant distribution ν . These results are different but complementary to the moment bounds on $W_1(\nu_n,\nu) - \mathbb{E}(W_1(\nu_n,\nu))$ obtained by Chazottes and Gouëzel [1] and Gouëzel and Melbourne [10] as a consequence of a concentration inequality for separately Lipschitz functionals of (T, T^2, \ldots, T^n) . See Section 4.3 for a deeper discussion.

All along the paper, the notation $a_n \ll b_n$ means that there exists a numerical constant C not depending on n such that $a_n \leq Cb_n$, for all positive integers n.

2 A Marcinkiewicz-Zygmund type inequality

Our first result extends Proposition 4 of Dedecker and Doukhan [3] to smooth Banach spaces belonging to $C_2(p, c_p)$.

Theorem 2.1. Let p be a real number in $[2, \infty[$ and let $(\mathbb{B}, |\cdot|_{\mathbb{B}})$ be a Banach space belonging to the class $C_2(p, c_p)$. Let $(X_i)_{i \in \mathbb{N}}$ be a sequence of centered random variables in $\mathbb{L}^p(\mathbb{B})$. Let $(\mathcal{F}_i)_{i\geq 0}$ be an increasing sequence of σ -algebras such that X_i is \mathcal{F}_i -measurable, and denote by $\mathbb{E}_i(\cdot) = \mathbb{E}(\cdot|\mathcal{F}_i)$ the conditional expectation with respect to \mathcal{F}_i . Define then

$$b_{i,n} = \max_{i \le \ell \le n} \left(\mathbb{E}_0 \left(|X_i|_{\mathbb{B}}^{p/2} \Big| \sum_{k=i}^{\ell} \mathbb{E}_i(X_k) \Big|_{\mathbb{B}}^{p/2} \right) \right)^{2/p}$$

For any integer $n \ge 0$, the following inequality holds:

$$\mathbb{E}_0(|S_n|_{\mathbb{B}}^p) \le K^p \Big(\sum_{i=1}^n b_{i,n}\Big)^{p/2} almost \ surely, \ where \ K = \sqrt{2p^{-1}} \sqrt{\max(c_p, p/2)} \,.$$
(2.1)

Remark 2.1. Taking $\mathcal{F}_0 = \{\Omega, \emptyset\}$, it follows that, for any integer $n \ge 0$,

$$\mathbb{E}(|S_n|_{\mathbb{B}}^p) \le K^p \Big(\sum_{i=1}^n \max_{i \le \ell \le n} \left\| |X_i|_{\mathbb{B}} \right\|_{k=i} \mathbb{E}(X_k|\mathcal{F}_i) \Big\|_{\mathbb{B}} \Big\|_{p/2} \Big)^{p/2} \text{ where } K = \sqrt{2p^{-1}} \sqrt{\max(c_p, p/2)}.$$

$$(2.2)$$

In addition, if we assume that $\mathbb{P}(|X_k|_{\mathbb{B}} \leq M) = 1$ for any $k \in \{1, \ldots, n\}$, inequality (2.2) combined with Proposition 5.2 of the appendix leads to the bound

$$\mathbb{E}\Big(\max_{1\le k\le n} |S_k|_{\mathbb{B}}^p\Big) \le C_p M^{p-1} n^{p/2} \Big(\sum_{k=0}^{n-1} \theta^{2/p}(k)\Big)^{p/2},$$
(2.3)

where

$$C_p = \frac{1}{2} \left(\frac{2pK}{p-1}\right)^p + 2^{3p-4} 3^p p \quad and \quad \theta(k) = \max\left\{ \mathbb{E}(|\mathbb{E}(X_i|\mathcal{F}_{i-k})|_{\mathbb{B}}), i \in \{k+1,\dots,n\} \right\}.$$

A complete proof of Inequality 2.3 will be given in Section 5.4.

When $\mathbb{B} = \mathbb{L}^q$ for $q \ge 2$, the constant K of Inequality (2.2) is equal to $\sqrt{2}(\max(p,q)-1)$. However we notice that we can obtain a better constant when the underlying sequence is a martingale differences sequence. More precisely, the following extension of the Marcinkiewicz-Zygmund type inequality obtained by Rio (2009) when the random variables are real-valued holds:

Theorem 2.2. Let p be a real number in $[2, \infty[$ and let $(\mathbb{B}, |\cdot|_{\mathbb{B}})$ be a Banach space belonging to the class $C_2(p, c_p)$. Let $(d_i)_{i \in \mathbb{N}}$ be a sequence of martingale differences with values in \mathbb{B} with respect to an increasing filtration $(\mathcal{F}_i)_{i \in \mathbb{N}}$ and such that for all $i \in \mathbb{N}$, $|||d_i|_{\mathbb{B}}||_p < \infty$. Then, setting $M_n = \sum_{i=1}^n d_i$, the following inequality holds:

$$\mathbb{E}(|M_n|_{\mathbb{B}}^p) \le (p^{-1}c_p)^{p/2} \Big(\sum_{i=1}^n ||d_i|_{\mathbb{B}}||_p^2\Big)^{p/2}.$$
(2.4)

The proof is omitted since it follows closely the lines of the proof of Proposition 2.1 in [19] (the bound (2.1) in [19] is obtained by using Inequality (1.3)).

In particular if $\mathbb{B} = \mathbb{L}^q(\mathcal{X}, \mathcal{A}, \mu)$ with $q \in [2, \infty[$ and (T, \mathcal{A}, ν) a measure space, Inequality (2.4) combined with Lemma 1.1 leads to

$$\mathbb{E}(|M_n|_q^p) \le (\max(p,q)-1)^{p/2} \Big(\sum_{i=1}^n |||d_i|_q||_p^2\Big)^{p/2},$$
(2.5)

 $|\cdot|_q$ being the norm on $\mathbb{L}^q(\mathcal{X}, \mathcal{A}, \mu)$.

Proof of Theorem 2.1. As in [18] and [3], we shall prove the result by induction. For any $t \in [0, 1]$ let

$$h_n(t) = \mathbb{E}_0(|S_{n-1} + tX_n|_{\mathbb{B}}^p).$$
(2.6)

Our induction hypothesis at step n-1 is the following: for any $k \leq n-1$,

$$h_k(t) \le K^p \Big(\sum_{i=1}^{k-1} b_{i,k} + t b_{k,k}\Big)^{p/2}.$$
 (2.7)

Since $K \ge 1$, the above inequality is clearly true for k = 1. Assuming that it is true for n - 1, let us prove it at step n.

Assume that one can prove that

$$h_n(t) \le \max(c_p, p/2) \Big(\sum_{k=1}^{n-1} b_{k,n} \int_0^1 (h_k(s))^{1-2/p} ds + b_{n,n} \int_0^t (h_n(s))^{1-2/p} ds \Big), \qquad (2.8)$$

then, using our induction hypothesis, it follows that

$$h_{n}(t) \leq \max(c_{p}, p/2) \Big(\sum_{k=1}^{n-1} b_{k,n} \int_{0}^{1} K^{p-2} \Big(\sum_{i=1}^{k-1} b_{i,k} + sb_{k,k} \Big)^{(p-2)/2} ds + b_{n,n} \int_{0}^{t} \big(h_{n}(s) \big)^{1-2/p} \Big) \\ \leq \max(c_{p}, p/2) \Big(K^{p-2} \sum_{k=1}^{n-1} b_{k,n} \int_{0}^{1} \big(\sum_{i=1}^{k-1} b_{i,n} + sb_{k,n} \big)^{(p-2)/2} ds + b_{n,n} \int_{0}^{t} \big(h_{n}(s) \big)^{1-2/p} ds \Big).$$

Integrating with respect to s, we get

$$b_{k,n} \int_0^1 \left(\sum_{i=1}^{k-1} b_{i,n} + sb_{k,n}\right)^{(p-2)/2} ds = \frac{2}{p} \left(\sum_{i=1}^k b_{i,n}\right)^{p/2} - \frac{2}{p} \left(\sum_{i=1}^{k-1} b_{i,n}\right)^{p/2}$$

implying that

$$\sum_{k=1}^{n-1} b_{k,n} \int_0^1 \left(\sum_{i=1}^{k-1} b_{i,n} + s b_{k,n} \right)^{(p-2)/2} ds = 2p^{-1} \left(\sum_{i=1}^{n-1} b_{i,n} \right)^{p/2}.$$

Therefore, since $K^2 = 2p^{-1} \max(c_p, p/2)$,

$$h_n(t) \le K^p \Big(\sum_{i=1}^{n-1} b_{i,n}\Big)^{p/2} + \max(c_p, p/2) b_{n,n} \int_0^t (h_n(s))^{1-2/p} ds \,.$$
(2.9)

Let $H_n(t) = \int_0^t (h_n(s))^{1-2/p} ds$. The differential integral inequation (2.9) writes

$$H'_{n}(s) \left(K^{p} \left(\sum_{i=1}^{n-1} b_{i,n} \right)^{p/2} + \max(c_{p}, p/2) b_{n,n} H(s) \right)^{-1+2/p} \le 1.$$

Setting

$$R_n(s) = \left(K^p \left(\sum_{i=1}^{n-1} b_{i,n}\right)^{p/2} + \max(c_p, p/2) b_{n,n} H(s)\right)^{2/p},$$

the previous inequality can be rewritten as

$$R'_n(s) \le 2p^{-1} \max(c_p, p/2) b_{n,n}$$

Integrating between 0 and t, we derive

$$(h_n(t))^{2/p} - K^2 \sum_{i=1}^{n-1} b_{i,n} \le R_n(t) - R_n(0) \le 2tp^{-1} \max(c_p, p/2)b_{n,n}.$$

Taking into account that $K^2 = 2p^{-1} \max(c_p, p/2)$, it follows that

$$(h_n(t))^{2/p} \le K^2 \left(\sum_{i=1}^{n-1} b_{i,n} + tb_{n,n}\right),$$

showing that our induction hypothesis holds true at step n. To end the proof it suffices to prove (2.8). We shall proceed as in the proof of Theorem 2.3 in [18]. With this aim, let

$$S_n(t) = \sum_{i=1}^n Y_i(t)$$
, where $Y_i(t) = X_i$ for $1 \le i \le n-1$ and $Y_n(t) = tX_n$.

Notice that for any integer k in [1, n-1], $S_k(t) = S_k$. Let now ψ_p be defined by (1.2). Applying second order Taylor expansion, we get

$$\psi_p(S_n(t)) = \sum_{i=1}^n \left(\psi_p(S_i(t)) - \psi_p(S_{i-1}(t)) \right)$$

= $\sum_{k=1}^n D\psi_p(S_{k-1})(Y_k(t)) + \sum_{i=1}^n \int_0^1 (1-s)D^2\psi_p(S_{i-1}+sY_i(t))(Y_i(t),Y_i(t))ds.$

But, for any integer k in [1, n],

$$D\psi_p(S_{k-1})(Y_k(t)) = \sum_{i=1}^{k-1} \left(D\psi_p(S_i)(Y_k(t)) - D\psi_p(S_{i-1})(Y_k(t)) \right)$$
$$= \sum_{i=1}^{k-1} \int_0^1 D^2 \psi_p(S_{i-1} + sX_i)(Y_k(t), X_i) ds \,.$$

Notice now that for any x and u in \mathbb{B} , $D^2\psi_p(x)(u,u) \ge 0$. Indeed, the function $x \mapsto \psi_p(x) = |x|_{\mathbb{B}}^{p/2}$ is convex for any $p \ge 2$ and is by assumption 2-times differentiable, implying that $D^2\psi_p(x)(u,u) \ge 0$. Therefore

$$\begin{split} \psi_p(S_n(t)) &\leq \sum_{i=1}^{n-1} \int_0^1 D^2 \psi_p(S_{i-1} + sX_i) \Big(\sum_{k=i+1}^n Y_k(t), X_i \Big) ds \\ &+ \sum_{i=1}^n \int_0^1 D^2 \psi_p(S_{i-1} + sY_i(t)) (Y_i(t), Y_i(t)) ds \,. \end{split}$$

Taking the conditional expectation w.r.t. \mathcal{F}_0 and recalling the definition (2.6) of $h_n(t)$, it follows that, for any $t \in [0, 1]$,

$$h_n(t) \le \sum_{i=1}^{n-1} \int_0^1 \mathbb{E}_0 \Big(D^2 \psi_p(S_{i-1} + sX_i) \Big(\sum_{k=i}^{n-1} X_k + tX_n, X_i \Big) ds \Big) + t^2 \int_0^1 \mathbb{E}_0 \Big(D^2 \psi_p(S_{n-1} + stX_n)(X_n, X_n) ds \Big).$$

Using again the fact that $D^2\psi_p(v)(u,u) \ge 0$, we have

$$t^{2} \int_{0}^{1} \mathbb{E}_{0} \Big(D^{2} \psi_{p} (S_{n-1} + stX_{n})(X_{n}, X_{n}) ds \Big) \leq \int_{0}^{t} \mathbb{E}_{0} \Big(D^{2} \psi_{p} (S_{n-1} + uX_{n})(X_{n}, X_{n}) du \Big) \,.$$

Hence setting

$$a_{i,n}(t) = X_i + \sum_{k=i+1}^{n-1} \mathbb{E}(X_k | \mathcal{F}_i) + t \mathbb{E}(X_n | \mathcal{F}_i),$$

and using the fact that (\mathcal{F}_i) is an increasing sequence of σ -algebras, we derive

$$h_n(t) \le \sum_{i=1}^{n-1} \int_0^1 \mathbb{E}_0 \Big(D^2 \psi_p(S_{i-1} + sX_i)(a_{i,n}(t), X_i) ds \Big) + \int_0^t \mathbb{E}_0 \Big(D^2 \psi_p(S_{n-1} + sX_n)(X_n, X_n) ds \Big) \,.$$

Notice now that since $(\mathbb{B}, |\cdot|_{\mathbb{B}})$ belongs to the class $\mathcal{C}_2(p, c_p)$ then (1.3) holds and Cauchy-Schwarz's inequality then implies that: for all x, u, v in \mathbb{B} ,

$$\left| D^2 \psi_p(x)(u,v) \right| \le c_p |x|_{\mathbb{B}}^{p-2} |u|_{\mathbb{B}} |v|_{\mathbb{B}}.$$

It follows that

$$h_n(t) \le c_p \sum_{i=1}^{n-1} \int_0^1 \mathbb{E}_0 \Big(|S_{i-1} + sX_i|_{\mathbb{B}}^{p-2} |a_{i,n}(t)|_{\mathbb{B}} |X_i|_{\mathbb{B}} \Big) ds + c_p \int_0^t \mathbb{E}_0 \Big(|S_{n-1} + sX_n|_{\mathbb{B}}^{p-2} |X_n|_{\mathbb{B}}^2 \Big) ds \,.$$

Hölder's inequality entails that

$$h_{n}(t) \leq c_{p} \sum_{i=1}^{n-1} \int_{0}^{1} (h_{i}(s))^{(p-2)/p} \Big(\mathbb{E}_{0} \big(|a_{i,n}(t)|_{\mathbb{B}}^{p/2} |X_{i}|_{\mathbb{B}}^{p/2} \big) \Big)^{2/p} ds + c_{p} \int_{0}^{t} \big(h_{n}(s) \big)^{(p-2)/p} \Big(\mathbb{E}_{0} (|X_{n}|_{\mathbb{B}}^{p}) \Big)^{2/p} ds .$$
(2.10)

Let $G_{i,n}(t) = \mathbb{E}_0(|a_{i,n}(t)|_{\mathbb{B}}^{p/2}|X_i|_{\mathbb{B}}^{p/2})$. Since it is a convex function, for any $t \in [0,1]$,

$$G_{i,n}(t) \le \max\left(G_{i,n}(0), G_{i,n}(1)\right) \le b_{i,n}^{p/2}.$$
 (2.11)

Starting from (2.10), using (2.11) and the fact that $(\mathbb{E}_0(|X_n|_{\mathbb{B}}^p))^{2/p} \leq b_{n,n}$, the inequality (2.8) follows. \diamond

3 Hoeffding type inequalities for martingales

In the following corollary, we give an exponential inequality for the deviation of the \mathbb{L}^{q} -norm of martingales.

Corollary 3.1. Let $q \in [2, \infty[$ and $(\mathcal{X}, \mathcal{A}, \mu)$ a measure space. Let $(d_i)_{i \in \mathbb{N}}$ be a sequence of martingale differences with values in $\mathbb{L}^q = \mathbb{L}^q(\mathcal{X}, \mathcal{A}, \mu)$ (equipped with the norm $|\cdot|_q$) with respect to an increasing filtration $(\mathcal{F}_i)_{i \in \mathbb{N}}$. Assume that for all $i \in \mathbb{N}$, there exists a positive real b such that $|||d_i|_q||_{\infty} \leq b$. Let $M_n = \sum_{i=1}^n d_i$. For any positive integer n and any positive real x, the following inequality holds

$$\mathbb{P}\Big(\max_{1 \le k \le n} |M_k|_q \ge x\Big) \le \begin{cases} 1 & \text{if } x < b\sqrt{(q-1)n} \\ \frac{(b^2(q-1)n)^{q/2}}{x^q} & \text{if } b\sqrt{(q-1)n} < x < b\sqrt{e(q-1)n} \\ \frac{1}{\sqrt{e}} \exp\left(-\frac{x^2}{2eb^2n}\right) & \text{if } x \ge b\sqrt{e(q-1)n} . \end{cases}$$
(3.1)

Under the assumptions of Corollary 3.1, Theorem 3.5 in [16] gives the following upper bound: for any positive integer n and any positive real x,

$$\mathbb{P}\left(\max_{1\le k\le n} |M_k|_q \ge x\right) \le 2\exp\left(-\frac{x^2}{2(q-1)b^2n}\right).$$
(3.2)

It is noteworthy to indicate that for any $q \ge e+1$, the bound in (3.1) is always better than the one given in (3.2). However, note that Theorem 3.5 in [16] holds not only in \mathbb{L}^q $(q \ge 2)$ but also in any 2-smooth Banach space.

Proof of Corollary 3.1. Let p be a real number in $[2, \infty[$. By the Doob-Kolmogorov maximal inequality,

$$\mathbb{P}\Big(\max_{1\leq k\leq n}|M_k|_q\geq x\Big)\leq x^{-p}\mathbb{E}\big(|M_n|_q^p\big)\,.$$

Therefore, using Inequality (2.5), we derive that

$$\mathbb{P}\left(\max_{1\leq k\leq n}|M_k|_q\geq x\right)\leq \left(\frac{\sqrt{a_pb^2n}}{x}\right)^p, \text{ where } a_p=\max(p,q)-1.$$

Taking p = q if $x < ((q-1)eb^2n)^{1/2}$ (so in this case $a_p = q-1$) and $p = 1 + \frac{x^2}{eb^2n}$ if $x \ge ((q-1)eb^2n)^{1/2}$ (so in this case $a_p = p-1$), the inequality (3.1) follows. \diamond

In the following corollary, we give an exponential inequality for the deviation of the \mathbb{L}^{q} -norm of partial sums. The proof is omitted since it is exactly the same as that of Corollary 3.1, by using Inequality (2.2) instead of Inequality (2.5).

Corollary 3.2. Let $q \in [2, \infty[$ and $(\mathcal{X}, \mathcal{A}, \mu)$ a measure space. Let $(X_i)_{i \in \mathbb{N}}$ be a sequence of random variables with values in $\mathbb{L}^q = \mathbb{L}^q(\mathcal{X}, \mathcal{A}, \mu)$ (equipped with the norm $|\cdot|_q$). Let $(\mathcal{F}_i)_{i\geq 0}$ be an increasing sequence of σ -algebras such that X_i is \mathcal{F}_i -measurable, and denote by $\mathbb{E}_i(\cdot) = \mathbb{E}(\cdot|\mathcal{F}_i)$ the conditional expectation with respect to \mathcal{F}_i . For any positive integer n, let $S_n = \sum_{i=1}^n X_i$. Assume that for any integer $i \in [1, n]$,

$$\left\| |X_i|_q \right\| \sum_{k=i}^n \mathbb{E}_i(X_k) \Big|_q \right\|_{\infty} \le b_n^2.$$

Then, for any positive real x, the following inequality holds

$$\mathbb{P}\Big(|S_n|_q \ge x\Big) \le \begin{cases} 1 & \text{if } x < b_n \sqrt{2(q-1)n} \\ \frac{(2b_n^2(q-1)n)^{q/2}}{x^q} & \text{if } b_n \sqrt{2(q-1)n} < x < b_n \sqrt{2e(q-1)n} \\ \frac{1}{\sqrt{e}} \exp\left(-\frac{x^2}{4eb_n^2 n}\right) & \text{if } x \ge b_n \sqrt{2e(q-1)n}. \end{cases}$$

4 Moment and deviation inequalities for the empirical process of nonuniformly expanding maps

In this section, we shall apply Theorem 2.1 and the inequalities recalled in the Appendix to obtain moment and deviation inequalities for the \mathbb{L}^q norm of the centered empirical distribution function of nonuniformly expanding maps of the interval. More precisely, our results apply to the iterates of a map T from [0, 1] to [0, 1] that can be modelled by a Young tower with polynomial tails of the return time.

In Section 4.1, we recall the formalism of Young towers, which has been described in many papers (see for instance [20] and [13]) with sometimes slight differences. Here we borrow the formalism described in Chapter 1 of Gouëzel's PhD thesis [8].

The moment inequalities are stated in Section 4.2, and an application to the Wassertein metric between the empirical measure of $\{T, T^2, \ldots, T^n\}$ and the *T*-invariant distribution is presented in Section 4.3. To be complete, we give in Section 4.4 some upper bounds for the maximum of the partial sums of Hölder observables, which can be proved as in Section 4.2.

4.1 One dimensional maps modelled by Young towers

Let T be a map from [0,1] to [0,1], and λ be a probability measure on [0,1]. Let Y be a Borel set of [0,1], with $\lambda(Y) > 0$. Assume that there exist a partition (up to a negligible set) $\{Y_k\}_{k \in \{1,...,K\}}$ of Y (note that K can be infinite) and a sequence $(\varphi_k)_{k \in \{1,...,K\}}$ of increasing numbers such that $T^{\varphi_k}(Y_k) = Y$. Let then φ_Y be the function from Y to $\{\varphi_k\}_{k \in \{1,...,K\}}$ such that $\varphi_Y(y) = \varphi_k$ if $y \in Y_k$.

We then define a space

$$X = \{(y, i) : y \in Y, i < \varphi_Y(y)\}$$

and a map \overline{T} on X:

$$\bar{T}(y,i) = \begin{cases} (y,i+1) & \text{if } i < \varphi_Y(y) - 1\\ (T^{\varphi_Y(y)}(y),0) & \text{if } i = \varphi_Y(y) - 1. \end{cases}$$

The space X is the Young tower. One can define the floors $\Delta_{k,i}$ for $k \in \{1, \ldots, K\}$ and $i \in \{0, \ldots, \varphi_k - 1\}$: $\Delta_{k,i} = \{(y,i) : y \in Y_k\}$. These floors define a partition of the tower:

$$X = \bigcup_{k \in \{1, \dots, K\}, i \in \{0, \dots, \varphi_k - 1\}} \Delta_{k, i}$$

On X, the measure m is defined as follows: if \overline{B} is a set included in $\Delta_{k,i}$, that can be written as $\overline{B} = B \times \{i\}$ with $B \subset Y_k$, then $m(\overline{B}) = \lambda(B)$. Consequently, for a set $\overline{A} \subset \bigcup_{\{k:\varphi_k > i\}} \Delta_{k,i}$, which can be written as $\overline{A} = A \times \{i\} = (\bigcup_{\{k:\varphi_k > i\}} B_k) \times \{i\}$ with $B_k \subset Y_k$, one has

$$m(\bar{A}) = \lambda(A) = \sum_{\{k \, : \, \varphi_k > i\}} \lambda(B_k)$$

Let π be the "projection" from X to [0,1] defined by $\pi(y,i) = T^i(y)$. Then, one has

$$\pi \circ \bar{T} = T \circ \pi \,.$$

Indeed, if $i < \varphi_Y(y) - 1$, then $\overline{T}(y, i) = (y, i + 1)$ so that

$$\pi \circ \overline{T}(y,i) = \pi(y,i+1) = T^{i+1}(y) = T \circ \pi(y,i)$$

If $i = \varphi_Y(y) - 1$, then $\overline{T}(y, i) = (T^{\varphi_Y(y)}(y), 0)$ so that

$$\pi \circ \bar{T}(y, \varphi_Y(y) - 1) = T^{\varphi_Y(y)}(y) = T(T^{\varphi_Y(y) - 1}(y)) = T \circ \pi(y, \varphi_Y(y) - 1).$$

Assume now that \overline{T} preserves the probability $\overline{\nu}$ on X, and let ν be the image measure of $\overline{\nu}$ by π . Then, for any measurable and bounded function f,

$$\nu(f(T)) = \bar{\nu}(f(T \circ \pi)) = \bar{\nu}((f \circ \pi)(\bar{T})) = \bar{\nu}(f \circ \pi) = \nu(f) + \bar{\nu}(f \circ \pi) = \bar{\nu}(f$$

and consequently ν is invariant by T.

The map T can be modelled by a Young tower if:

1. For any $k \in \{1, ..., K\}$, T^{φ_k} is a measurable isomorphism between Y_k and Y. Moreover there exists C > 0 such that, for any $k \in \{1, ..., K\}$ and almost every x, y in Y_k ,

$$\left|1 - \frac{(T^{\varphi_k})'(x)}{(T^{\varphi_k})'(y)}\right| \le C |T^{\varphi_k}(x) - T^{\varphi_k}(y)|$$

2. There exists C > 0 such that, for any $k \in \{1, ..., K\}$ and almost every x, y in Y_k , for any $i < \varphi_k$,

$$|T^{i}(x) - T^{i}(y)| \le C|T^{\varphi_{k}}(x) - T^{\varphi_{k}}(y)|$$

3. There exists $\tau > 1$ such that, for any $k \in \{1, \ldots, K\}$ and almost every x, y in Y_k :

$$|T^{\varphi_k}(x) - T^{\varphi_k}(y)| \ge \tau |x - y|.$$

4. $\sum_{k=1}^{K} \varphi_k \lambda(Y_k) < \infty$.

If T can be modelled by a Young tower, then, on the tower, there exists a unique \overline{T} -invariant probability measure $\overline{\nu}$ which is absolutely continuous with respect to m. Hence, there exists a unique T-invariant measure ν which is absolutely continuous with respect to the measure λ (see [8], Proposition 1.3.18). This measure is the image measure of $\overline{\nu}$ by the projection π and is supported by

$$\Lambda = \bigcup_{n \ge 0} T^n(Y).$$

Let \bar{Y} be the basis of the tower, that is $\bar{Y} = \{(y,0), y \in Y\}$. Let $\varphi_{\bar{Y}}$ be the function from \bar{Y} to $\{\varphi_k\}_{k \in \{1,\dots,K\}}$ such that $\varphi_{\bar{Y}}((y,0)) = \varphi_Y(y)$. By definition of \bar{T} one gets $\bar{T}^{\varphi_k}(\Delta_{k,0}) = \bar{Y}$. In addition, the quantity $\bar{\nu}(\{(y,0) \in \bar{Y} : \varphi_{\bar{Y}}((y,0)) > k\})$ is exactly of the same order as $\lambda(\{y \in Y : \varphi_Y(y) > k\})$ (see [8], Proposition 1.1.24).

On the tower, one defines the distance s as follows: s(x, y) = 0 is x and y do not belong to the same partition element $\Delta_{k,i}$. If x = (a, i) and y = (b, i) belong to the same $\Delta_{k,i}$ (meaning that a and b belong to Y_k), then $\delta(x, y) = \beta^{s(x,y)}$ for $\beta = 1/\tau$, where s(x, y) is the smallest integer n such that $S^n(a)$ and $S^n(b)$ are not in the same Y_j .

Because of Item 3, we know that $|S'| \ge \tau > 1$, so that S is uniformly expanding. For x = (a, i) and y = (b, i) in $\Delta_{k,i}$, one has

$$|\pi(x) - \pi(y)| = |T^{i}(a) - T^{i}(b)| \le C|T^{\varphi_{k}}(a) - T^{\varphi_{k}}(b)|$$

by Item 2. Since $T^{\varphi_k} = S$ on Y_k , and since $|S'| \ge \tau$, it follows that

$$|\pi(x) - \pi(y)| \le C\beta^{s(x,y)-1} \le \frac{C}{\beta}\beta^{s(x,y)}$$

Now, if x and y do not belong to the same partition element $\Delta_{k,i}$, then $|\pi(x) - \pi(y)| \leq \beta^{s(x,y)} = 1$. It follows that there exists a positive constant K such that

$$|\pi(x) - \pi(y)| \le K\beta^{s(x,y)},$$

meaning that π is Lipschitz with respect to the distance δ .

Among the maps that can be modelled by a Young tower, we shall consider the maps defined as follows.

Definition 4.1. One says that the map T can be modelled by a Young tower with polynomial tails of the return times of order $1/\gamma$ with $\gamma \in (0,1)$ if $\lambda(\{y \in Y : \varphi_Y(y) > k\}) \leq Ck^{-1/\gamma}$.

Let us briefly describe some properties of such maps. For $\alpha \in (0, 1]$, let $\delta_{\alpha} = \delta^{\alpha}$, let L_{α} be the space of Lipschitz functions with respect to δ_{α} , and let

$$L_{\alpha}(f) = \sup_{x,y \in X} \frac{|f(x) - f(y)|}{\delta_{\alpha}(x,y)}.$$
(4.1)

For any positive real a, let $L_{\alpha,a}$ be the set of functions such that $L_{\alpha}(f) \leq a$.

Denote by P the Perron-Frobenius operator of T with respect to $\bar{\nu}$: for any bounded measurable functions φ, ψ ,

$$\bar{\nu}(\varphi \cdot \psi \circ \bar{T}) = \bar{\nu}(P(\varphi)\psi).$$

Let T be a map that can be modelled by a Young tower with polynomial tails of the return times of order $1/\gamma$. Then one can prove that (see [13] and Lemma 2.2 in [7]): for any $m \ge 1$ and any $\alpha \in (0, 1]$, there exists $C_{\alpha} > 0$ such that, for any $\psi \in L_{\alpha}$,

$$|P^{m}(\psi)(x) - P^{m}(\psi)(y)| \le C_{\alpha}\delta_{\alpha}(x,y)L_{\alpha}(\psi).$$

$$(4.2)$$

Moreover, starting from the results by Gouëzel [8], we shall prove in Proposition 5.3 of the appendix that, for any $\alpha \in (0, 1]$ there exists $K_{\alpha} > 0$ such that

$$\bar{\nu}\Big(\sup_{f\in L_{\alpha,1}}|P^n(f)-\bar{\nu}(f)|\Big) \le \frac{K_{\alpha}}{n^{(1-\gamma)/\gamma}}.$$
(4.3)

A well known example of map which can be modelled by a Young tower with polynomial tails of the return times is the intermittent map T_{γ} introduced by Liverani *et al.* [12]: for $\gamma \in (0, 1)$,

$$T_{\gamma}(x) = \begin{cases} x(1+2^{\gamma}x^{\gamma}) & \text{if } x \in [0,1/2[\\ 2x-1 & \text{if } x \in [1/2,1]; \end{cases}$$
(4.4)

For this map, λ is the Lebesgue measure on [0, 1] and one can take Y = [1/2, 1]. Let $x_0 = 1$, and define recursively $x_{n+1} = T_{\gamma}^{-1}(x_n) \cap [0, 1/2]$. One can prove that $x_n = \frac{1}{2}(\gamma n)^{-1/\gamma}$. Let then $y_n = T_{\gamma}^{-1}(x_{n-1}) \cap [1/2, 1]$. The y_k 's are built in such a way that $Y_k =]y_{k+1}, y_k]$ is the set of points y in Y for which $T_{\gamma}^k(Y_k) = Y$. One can verify, by controling explicitly the distortion, that the items 1,2 and 3 are satisfied with $\varphi_k = k$. Item 4 follows from the fact that $\sum_{k=1}^{\infty} k\lambda(Y_k) \leq C \sum_{k=1}^{\infty} kk^{-(\gamma+1)/\gamma} < \infty$, since $\gamma \in (0, 1)$. Moreover, one has

$$\lambda(\{y \in Y : \varphi_Y(y) > k\}) = \sum_{i=k+1}^{\infty} \lambda(Y_i) \le Ck^{-1/\gamma},$$

so that the tail of the return times is of order $1/\gamma$.

4.2 Moment and deviation inequalities for the empirical process

For any $q \in [2, \infty]$, let

$$D_{n,q} = \left(\int_0^1 |G_n(t)|^q dt\right)^{1/q},$$
(4.5)

where G_n is defined by

$$G_n(t) = \sum_{k=1}^n \left(\mathbf{1}_{T^k \le t} - \nu([0, t]) \right), \ t \in [0, 1].$$
(4.6)

By duality, it is well-known that

$$\frac{1}{n}D_{n,q} = \sup_{f \in W_{q',1}} \left| \frac{1}{n} \sum_{k=1}^{n} \left(f(T^k) - \nu(f) \right) \right|, \tag{4.7}$$

where q' = q/(q-1) and $W_{q',1}$ is the Sobolev ball

$$W_{q',1} = \left\{ f : \int_0^1 |f'(x)|^{q'} dx \le 1 \right\}.$$
(4.8)

Consequently, a moment inequality on $D_{n,q}$ provides a concentration inequality of the empirical measure of $\{T, T^2, \dots, T^n\}$ around ν , on a class of smooth functions. Note that, the class $W_{q',1}$ is larger as q increases, and always contains the class of Lipschitz functions with Lipschitz constant 1.

In what follows, we shall denote by $\|\cdot\|_{p,\nu}$ the \mathbb{L}^p -norm on $([0,1],\nu)$

Theorem 4.1. Let T be a map that can be modelled by a Young tower with polynomial tails of the return times of order $1/\gamma$ with $\gamma \in (0, 1/2)$, and let $p_{\gamma} = 2(1-\gamma)/\gamma$. For $q \in [2, \infty[$ let $D_{n,q}$ be defined by (4.5). Then, there exists a positive constant C such that for any $n \ge 1$,

$$\left\| \max_{1 \le k \le n} D_{k,q} \right\|_{p_{\gamma},\nu} \le C\sqrt{n}$$

As a consequence of Theorem 4.1, for any $\gamma \in (0, 1/2)$ and any positive real x,

$$u\left(\max_{1\leq k\leq n} D_{k,q} \geq x\sqrt{n}\right) \leq \frac{C}{x^{2(1-\gamma)/\gamma}}.$$

In addition, proceeding as at the beginning of page 872 of the paper [1], we infer that, under the assumptions of Theorem 4.1, for any real $p > 2(1 - \gamma)/\gamma$, there exists a positive constant C such that, for any $n \ge 1$,

$$\left\| \max_{1 \le k \le n} D_{k,q} \right\|_{p,\nu} \le C n^{(\gamma p + \gamma - 1)/(\gamma p)}.$$

Let us examine now the case where $\gamma \geq 1/2$.

Theorem 4.2. Let T be a map that can be modelled by a Young tower with polynomial tails of the return times of order $1/\gamma$ with $\gamma \in [1/2, 1)$. For $q \in [2, \infty[$, let $D_{n,q}$ be defined by (4.5).

1. There exists a positive constant C such that for any $n \ge 1$,

$$\left\| \max_{1 \le k \le n} D_{n,q} \right\|_{1/\gamma,\nu} \le C(n \log n)^{\gamma}$$

2. If $p > 1/\gamma$, then there exists a positive constant C such that for any $n \ge 1$,

$$\left\| \max_{1 \le k \le n} D_{k,q} \right\|_{p,\nu} \le C n^{(\gamma p + \gamma - 1)/(\gamma p)}$$

For the optimality of the moment bounds of Theorems 4.1 and 4.2, we refer the paper by Melbourne and Nicol [14] and to the recent paper by Gouëzel and Melbourne [10]. Since, for $q \geq 2$, the class $W_{q',1}$ contains the class of Lipschitz functions with Lipschitz constant 1, one can apply Proposition 1.1 and 1.2 in [10], showing that these bounds are optimal. See also Remark 4.1 below for more comments about the optimality.

Theorem 4.3. Let T be a map that can be modelled by a Young tower with polynomial tails of the return times of order $1/\gamma$ with $\gamma \in (1/2, 1)$. For $q \in [2, \infty[$, let $D_{n,q}$ be defined by (4.5). Then, there exists a positive constant C such that for any $n \ge 1$ and any positive real x,

$$\nu\Big(\max_{1\le k\le n} D_{k,q} \ge x \, n^{\gamma}\Big) \le C x^{-1/\gamma} \,. \tag{4.9}$$

Applying Theorem 4.3, one gets for $p \in [1, 1/\gamma[,$

$$\left\| \max_{1 \le k \le n} D_{k,q} \right\|_{p,\nu}^p = p \int_0^\infty x^{p-1} \nu \Big(\max_{1 \le k \le n} D_{k,q} \ge x \Big) dx \le p \int_0^{n^\gamma} x^{p-1} dx + Cnp \int_{n^\gamma}^\infty \frac{1}{x^{1+\gamma^{-1}-p}} dx \,.$$

Consequently, for $p \in [1, 1/\gamma]$, there exists a positive constant C such that

$$\left\| \max_{1 \le k \le n} D_{k,q} \right\|_{p,\nu} \le C n^{\gamma}.$$

Remark 4.1. Inequality (4.9) cannot hold for $\gamma = 1/2$. Indeed, for the map T_{γ} defined in (4.4), Item 1 of Theorem 1.1 in [2] implies that, for any positive real x,

$$\lim_{n \to \infty} \nu \left(\frac{1}{\sqrt{n \log n}} D_{n,2} > x \right) = \mathbb{P}(|N| > x) > 0,$$

where N is a real-valued centered Gaussian random variable with positive variance. In addition, for $\gamma \in (1/2, 1)$, Item 2 of the same paper implies that

$$\lim_{n \to \infty} \nu \left(\frac{1}{n^{\gamma}} D_{n,2} > t \right) = \mathbb{P}(|Z_{\gamma}| > t) > 0,$$

where Z_{γ} is an $1/\gamma$ -stable random variable such that $\lim_{x\to\infty} x^{1/\gamma} \mathbb{P}(|Z_{\gamma}| > x) = c > 0$.

4.3 Application to the Wasserstein metric between the empirical measure and the invariant measure

Let us give an application of the results of Section 4.2 to the Wasserstein distance between the empirical measure of $\{T, T^2, \ldots, T^n\}$ and the invariant distribution ν . Recall that Wasserstein distance W_1 between two probability measures ν_1 and ν_2 on [0, 1] is defined as

$$W_1(\nu_1, \nu_2) = \inf \left\{ \int |x - y| \mu(dx, dy), \mu \in \mathcal{M}(\nu_1, \nu_2) \right\}$$

where $\mathcal{M}(\nu_1, \nu_2)$ is the set of probability measures on $[0, 1] \times [0, 1]$ with margins ν_1 and ν_2 . Recall also that, in this one dimensional setting,

$$W_1(\nu_1,\nu_2) = \int_0^1 |F_{\nu_1}(t) - F_{\nu_2}(t)| dt \,,$$

where F_{ν_1} and F_{ν_2} are the distribution functions of ν_1 and ν_2 respectively. Therefore, setting

$$\nu_n = \frac{1}{n} \sum_{i=1}^n \delta_{T^i} \,,$$

we get that for any $q \ge 2$,

$$W_1(\nu_n,\nu) \le \frac{1}{n} D_{n,q}$$

The following corollary is a direct consequence of the results of Section 4.2.

Corollary 4.1. Let T be a map that can be modelled by a Young tower with polynomial tails of the return times of order $1/\gamma$ with $\gamma \in (0, 1)$.

- 1. If $\gamma \in (0, 1/2)$, then $||W_1(\nu_n, \nu)||_{p,\nu}^p \ll n^{-(1-\gamma)/\gamma}$ for any $p \ge 2(1-\gamma)/\gamma$.
- 2. If $\gamma \in [1/2, 1)$, then

$$\|W_1(\nu_n,\nu)\|_{p,\nu}^p \ll \begin{cases} n^{-(1-\gamma)/\gamma} \log n & \text{if } p = 1/\gamma \\ n^{-(1-\gamma)/\gamma} & \text{if } p > 1/\gamma. \end{cases}$$

3. If $\gamma \in (1/2, 1)$, then, for any $n \ge 1$ and any positive real x,

$$\nu(W_1(\nu_n,\nu) \ge x n^{\gamma-1}) \ll x^{-1/\gamma}.$$

In their Theorem 1.4, Gouëzel and Melbourne [10] obtain general bounds for the moment of separately Lipschitz functionals of (T, T^2, \ldots, T^n) , where T is a (non necessarily onedimensional) map that can be modelled by a Young tower with polynomial tails of the return times.

As a consequence of their results, one gets the same inequalities as in Corollary 4.1 but for the quantity $W_1(\nu_n,\nu) - \mathbb{E}(W_1(\mu_n,\nu))$ instead of $W_1(\nu_n,\nu)$. Note that the upper bounds for $W_1(\nu_n,\nu) - \mathbb{E}(W_1(\mu_n,\nu))$ are valid if T is nonuniformly expanding from \mathcal{X} to \mathcal{X} , where \mathcal{X} can be any metric space.

The two results are not of the same nature. However, in our one dimensional setting, the moments bounds of Corollary 4.1 imply the same moment bounds for $W_1(\nu_n, \nu) - \mathbb{E}(W_1(\mu_n, \nu))$, because $(\mathbb{E}(W_1(\nu_n, \nu)))^p \leq ||W_1(\mu_n, \nu)||_p^p$. The same remark does not hold for the deviation bounds, which are not directly comparable.

To conclude this section, let us mention that there is no hope to extend Corollary 4.1 to higher dimension with the same bounds. To see this, let us consider the case of \mathbb{R}^d -valued random variables (X_1, X_2, \ldots, X_n) that are bounded, independent, and identically distributed. Let ν_n be the empirical measure of $\{X_1, X_2, \ldots, X_n\}$ and ν be the common distribution of the X_i 's. It is well known that, when $d \geq 3$ and ν has a component which is absolutely continuous with respect to the Lebesgue measure, $\mathbb{E}(W_1(\nu_n, \nu))$ is exactly of order $n^{-1/d}$, which is much slower than $n^{-1/2}$.

4.4 Moment and deviation inequalities for partial sums

In Theorem 4.4 below, we assume that T is a nonuniformly expanding map on (\mathcal{X}, λ) with λ a probability measure on \mathcal{X} , and that T can be modelled by a Young tower. Contrary to the previous sections, \mathcal{X} can be any bounded metric space and not necessarily the unit interval. Let f be a Hölder continuous function from \mathcal{X} to \mathbb{R} and $S_n(f) = \sum_{i=1}^n (f \circ T^i - \nu(f))$.

Theorem 4.4. Let T be a map that can be modelled by a Young tower with polynomial tails of the return times of order $1/\gamma$ with $\gamma \in (0, 1)$.

- 1. If $\gamma \in (0, 1/2)$ then $\left\| \max_{1 \le k \le n} |S_k(f)| \right\|_{p, \nu}^p \ll n^{p (1 \gamma)/\gamma}$ for any $p \ge 2(1 \gamma)/\gamma$.
- 2. If $\gamma \in [1/2, 1)$, then

$$\left\|\max_{1\leq k\leq n} |S_k(f)|\right\|_{p,\nu}^p \ll \begin{cases} n\log n & \text{if } p=1/\gamma\\ n^{p-(1-\gamma)/\gamma} & \text{if } p>1/\gamma \end{cases}.$$

3. If $\gamma \in (1/2, 1)$, for any $n \ge 1$ and any positive real x,

$$\nu\Big(\max_{1\leq k\leq n}|S_k(f)|\geq x\,n^\gamma\Big)\ll x^{-1/\gamma}\,.$$

The proof is omitted since it is a simpler version of the proofs of Theorems 4.1, 4.2 and 4.3. Indeed the norm $|\cdot|_q$ is replaced by the absolute values and we do not need to deal with the supremum over a subset of the class of Hölder functions of order 1/q.

After this paper was written, we became aware that, using different methods based on martingale approximations, Gouëzel and Melbourne [10] had independently obtained the upper bounds given in Theorem 4.4 (but for $|S_n(f)|$ instead of $\max_{1 \le k \le n} |S_k(f)|$).

As in Section 4.2, applying Propositions 1.1 and 1.2 in [10], we see that the moments bounds of Theorem 4.4 cannot be improved.

Note also that, for the map T_{γ} defined in (4.4), we can make a similar remark as Remark 4.1: Firstly, Inequality (4.9) cannot hold for $\gamma = 1/2$. Indeed by Item 3 page 88 [9], if $f(0) \neq \nu(f)$, for any positive real x,

$$\lim_{n \to \infty} \nu \left(\frac{1}{\sqrt{n \log n}} |S_n(f)| > x \right) = \mathbb{P}(|N| > x) > 0,$$

where N is a real-valued centered Gaussian random variable with positive variance. In addition, for $\gamma \in (1/2, 1)$, Theorem 1.3 of the same paper implies that

$$\lim_{n \to \infty} \nu \left(|S_n(f)| > x n^{\gamma} \right) = \mathbb{P}(|Z_{\gamma}| > x) > 0 \,,$$

where Z_{γ} is an $1/\gamma$ -stable random variable such that $\lim_{x\to\infty} x^{1/\gamma} \mathbb{P}(|Z_{\gamma}| > x) = c > 0$.

For the intermittent map T_{γ} defined in (4.4), Theorem 4.4 also holds for observables with bounded variation (BV). More generally, Theorem 4.5 below shows that the conclusions of Theorem 4.4 also hold when we consider BV observables of the iterates of T_{γ} , where T_{γ} is a generalized Pomeau-Manneville map (or GPM map) of parameter $\gamma \in (0, 1)$ as defined in [4].

Theorem 4.5. Let T_{γ} be a GPM map of parameter $\gamma \in (0, 1)$ on the unit interval, with invariant measure ν_{γ} . Let f be a BV function from [0, 1] to \mathbb{R} . Then Items 1,2 and 3 of Theorem 4.4 hold for $S_n(f) = \sum_{i=1}^n (f \circ T_{\gamma}^i - \nu_{\gamma}(f))$.

4.5 Proofs of Theorems 4.1, 4.2, 4.3 and 4.5

Proof of Theorem 4.1. For any t, let f_t be the function defined by $f_t(x) = \mathbf{1}_{x \leq t}$. Notice first that, for any $p \geq 1$,

$$\begin{aligned} \left\| \max_{1 \le k \le n} D_{k,q} \right\|_{p,\nu}^{p} &= \nu \Big(\max_{1 \le k \le n} \Big| \int_{0}^{1} \Big| \sum_{i=1}^{k} (\mathbf{1}_{T^{i} \le t}) - \nu([0,t]) \Big|^{q} dt \Big|^{p/q} \Big) \\ &= \bar{\nu} \Big(\max_{1 \le k \le n} \Big| \int_{0}^{1} \Big| \sum_{i=1}^{k} (f_{t} \circ T^{i} \circ \pi - \bar{\nu}(f_{t} \circ \pi)) \Big|^{q} dt \Big|^{p/q} \Big) \\ &= \bar{\nu} \Big(\max_{1 \le k \le n} \Big| \int_{0}^{1} \Big| \sum_{i=1}^{k} (f_{t} \circ \pi \circ \bar{T}^{i} - \bar{\nu}(f_{t} \circ \pi)) \Big|^{q} dt \Big|^{p/q} \Big). \end{aligned}$$

Let $g_t := f_t \circ \pi$ and $G(x) = \{g_t(x), t \in [0, 1]\}$. Denote by $|\cdot|_q$ the norm associated to the Banach space $\mathbb{B} = \mathbb{L}^q([0, 1], dt)$. With these notations, we then have

$$\left\| \max_{1 \le k \le n} D_{k,q} \right\|_{p,\nu}^{p} = \bar{\nu} \left(\max_{1 \le k \le n} \left| \sum_{i=1}^{k} (G(\bar{T}^{i}) - \bar{\nu}(G(\bar{T}^{i}))) \right|_{q}^{p} \right).$$
(4.10)

Let now $(X_i)_{i\in\mathbb{N}}$ be a stationary Markov chain defined on a probability space $(\Omega, \mathcal{A}, \mathbb{P})$, with state space X, transition probability P and invariant distribution $\bar{\nu}$. Recall then (see for instance Lemma XI.3 [11]) that for every $n \geq 1$, we have the following equalities in law (where in the left-hand side the law is meant under $\bar{\nu}$ and in the right-hand side the law is meant under \mathbb{P})

$$(\bar{T}^n, \dots, \bar{T}) \stackrel{d}{=} (X_1, \dots, X_n)$$

$$\max_{1 \le k \le n} \Big| \sum_{i=1}^k (G(\bar{T}^i) - \bar{\nu}(G(\bar{T}^i))) \Big|_q \stackrel{d}{=} \max_{1 \le k \le n} \Big| \sum_{i=k}^n (G(X_i) - \mathbb{E}(G(X_i))) \Big|_q^p.$$
(4.11)

Therefore, starting from (4.10) and using (4.11), we infer that for any real $p \in [1, \infty[$,

$$\left\| \max_{1 \le k \le n} D_{k,q} \right\|_{p,\nu}^{p} = \mathbb{E} \left(\max_{1 \le k \le n} \left| \sum_{i=k}^{n} (G(X_{i}) - \mathbb{E}(G(X_{i}))) \right|_{q}^{p} \right) \\ \le 2^{p} \mathbb{E} \left(\max_{1 \le k \le n} \left| \sum_{i=1}^{k} (G(X_{i}) - \mathbb{E}(G(X_{i}))) \right|_{q}^{p} \right).$$
(4.12)

Whence, Theorem 4.1 will follow if one can prove that there exists a positive constant C such that for any $n \ge 1$,

$$\mathbb{E}\Big(\max_{1\leq k\leq n}\Big|\sum_{i=1}^{k} (G(X_i) - \mathbb{E}(G(X_i)))\Big|_{q}^{\frac{2(1-\gamma)}{\gamma}}\Big) \leq Cn^{\frac{1-\gamma}{\gamma}}.$$
(4.13)

With this aim, we shall apply the Rosenthal type inequality (5.2) given in Appendix, with $p = 2(1 - \gamma)/\gamma$ (note that p > 2 since $\gamma \in (0, 1/2)$). Letting $\mathcal{F}_k = \sigma(X_i, i \leq k)$ and $G^{(0)} = G - \mathbb{E}(G(X_1))$, this leads to

$$\mathbb{E}\Big(\max_{1\leq k\leq n}\Big|\sum_{i=1}^{k} (G(X_i) - \mathbb{E}(G(X_i)))\Big|_{q}^{\frac{2(1-\gamma)}{\gamma}}\Big) \ll n\mathbb{E}\Big(|G(X_1)|_{q}^{\frac{2(1-\gamma)}{\gamma}}\Big) \\
+ n\Big(\sum_{k=1}^{n} \frac{1}{k^{1+\delta\gamma/(1-\gamma)}}\Big\|\mathbb{E}_0\Big(\Big|\sum_{i=1}^{k} G^{(0)}(X_i)\Big|_{q}^{2}\Big)\Big\|_{(1-\gamma)/\gamma}^{\delta}\Big)^{\frac{(1-\gamma)}{\delta\gamma}}. \quad (4.14)$$

where $\delta = \min(1/2, \gamma/(2-4\gamma))$. To handle the terms $\|\mathbb{E}_0(|\sum_{i=1}^k G^{(0)}(X_i)|_q^2)\|_{(1-\gamma)/\gamma}$ in Inequality (4.14), we shall use Inequality (2.2) which together with Item 1 of Lemma 1.1 leads to

$$\mathbb{E}_{0}\left(\left|\sum_{i=1}^{k} G^{(0)}(X_{i})\right|_{q}^{2}\right) \leq 2(2q-3)\sum_{i=1}^{k}\sum_{\ell=i}^{k} \mathbb{E}_{0}(|G^{(0)}(X_{i})|_{q}|\mathbb{E}_{i}(G^{(0)}(X_{\ell}))|_{q})$$
$$\leq 2(2q-3)\sum_{i=1}^{k}\sum_{\ell=i}^{k} \mathbb{E}_{0}(|\mathbb{E}_{i}(G^{(0)}(X_{\ell}))|_{q}),$$

where for the last inequality, we have used the fact that for any i, $|G^{(0)}(X_i)|_q \leq 1$ almost surely. Hence

$$\left\| \mathbb{E}_{0} \left(\left| \sum_{i=1}^{k} G^{(0)}(X_{i}) \right|_{q}^{2} \right) \right\|_{(1-\gamma)/\gamma} \leq 2(2q-3) \sum_{i=1}^{k} \sum_{\ell=i}^{k} \| \mathbb{E}_{0}(|\mathbb{E}_{i}(G^{(0)}(X_{\ell}))|_{q}) \|_{(1-\gamma)/\gamma}.$$
(4.15)

Let us now handle the term $\|\mathbb{E}_0(|\mathbb{E}_i(G^{(0)}(X_\ell))|_q)\|_{(1-\gamma)/\gamma}$ in Inequality (4.15). With this aim, we first notice that

$$|\mathbb{E}_{i}(G^{(0)}(X_{\ell}))|_{q}^{q} = \int_{0}^{1} |\mathbb{E}(\mathbf{1}_{\pi(X_{\ell})\leq t}|X_{i}) - \mathbb{E}(\mathbf{1}_{\pi(X_{\ell})\leq t})|^{q} dt.$$

Again by a duality argument (as to prove (4.7)), we have

$$\int_0^1 \left| \mathbb{E}(\mathbf{1}_{\pi(X_\ell) \le t} | X_i) - \mathbb{E}(\mathbf{1}_{\pi(X_\ell) \le t}) \right|^q dt = \sup_{h \in W_{q',1}} \left| P_{\pi(X_\ell) | X_i}(h) - P_{\pi(X_\ell)}(h) \right|^q,$$

where the Sobolev ball $W_{q',1}$ is defined in (4.8), $P_{\pi(X_{\ell})|X_i}$ is the conditional distribution of $\pi(X_{\ell})$ given X_i , and $P_{\pi(X_{\ell})}$ is the distribution of $\pi(X_{\ell})$. Therefore

$$|\mathbb{E}_{i}(G^{(0)}(X_{\ell}))|_{q} = \sup_{h \in W_{q',1}} \left| P_{\pi(X_{\ell})|X_{i}}(h) - P_{\pi(X_{\ell})}(h) \right| = \sup_{h \in W_{q',1}} \left| P_{X_{\ell}|X_{i}}(h \circ \pi) - P_{X_{\ell}}(h \circ \pi) \right|,$$

where $P_{X_{\ell}|X_i}$ is the conditional distribution of X_{ℓ} given X_i , and $P_{X_{\ell}}$ is the distribution of X_{ℓ} . Notice now that if $f \in W_{q',1}$ then for any x and y in [0, 1],

$$|f(x) - f(y)| = \left| \int_{x}^{y} f'(t) dt \right| \le |x - y|^{1/q} \left(\int_{0}^{1} |f'(x)|^{q'} dx \right)^{1/q'}$$

Therefore,

$$W_{q',1} \subset H_{1/q,1} \,,$$

where $H_{1/q,1}$ is the set of functions that are 1/q-Hölder with Hölder constant 1. It follows that, for any $h \in W_{q',1}$, there exists a positive constant C such that

$$|h \circ \pi(x) - h \circ \pi(y)| \le |\pi(x) - \pi(y)|^{1/q} \le C\delta_{1/q}(x,y),$$

proving that $h \circ \pi$ belongs to the set $L_{1/q,C}$ defined right after (4.1). Let now

$$f_{\ell-i,h}(x) := |P_{X_{\ell}|X_i=x}(h \circ \pi) - P_{X_{\ell}}(h \circ \pi)| = |P^{\ell-i}(h \circ \pi)(x) - \bar{\nu}(h \circ \pi)|.$$

Using the triangle inequality, we have

$$|f_{\ell-i,h}(x) - f_{\ell-i,h}(y)| \le |P^{\ell-i}(h \circ \pi)(x) - P^{\ell-i}(h \circ \pi)(y)|.$$

Since $h \circ \pi$ belongs to $L_{1/q,C}$, the contraction property (4.2) entails that

$$|f_{\ell-i,h}(x) - f_{\ell-i,h}(y)| \le CC_{1/q}\delta_{1/q}(x,y).$$

Let $\widetilde{C} = CC_{1/q}$. We have shown that, for any $h \in W_{q',1}$, $f_{\ell-i,h} \in \mathcal{F}_{\ell-i} \subset L_{1/q,\widetilde{C}}$. Then, setting

$$m_{\ell-i}(x) = \sup_{h \in W_{q',1}} f_{\ell-i,h}(x)$$

we have $m_{\ell-i}(x) = \sup_{g \in \mathcal{F}_{\ell-i}} g(x)$. Therefore, if $m_{\ell-i}(x) \ge m_{\ell-i}(y)$,

$$m_{\ell-i}(x) - m_{\ell-i}(y) = g_x(x) - g_y(y) \le g_x(x) - g_x(y) \le \tilde{C}\delta_{1/q}(x,y),$$

since $\mathcal{F}_{\ell-i} \subset L_{1/q,\widetilde{C}}$. So overall,

$$|\mathbb{E}_{i}(G^{(0)}(X_{\ell}))|_{q} - \mathbb{E}|\mathbb{E}_{i}(G^{(0)}(X_{\ell}))|_{q} = m_{\ell-i}(X_{i}) - \mathbb{E}(m_{\ell-i}(X_{i})),$$

with $m_{\ell-i} \in L_{1/q,\widetilde{C}}$. Next, using (4.3), it follows that there exists a positive constant C such that, for any $i \ge 1$,

$$\|\mathbb{E}_0(|\mathbb{E}_i(G^{(0)}(X_\ell))|_q) - \mathbb{E}|\mathbb{E}_i(G^{(0)}(X_\ell))|_q\|_1 = \|P^i(m_{\ell-i}) - \bar{\nu}(m_{\ell-i})\|_1 \le Ci^{-(1-\gamma)/\gamma}.$$
 (4.16)

Using similar arguments we infer that there exists a positive constant C such that, for any $\ell \geq i+1$,

$$\||\mathbb{E}_{i}(G^{(0)}(X_{\ell}))|_{q}\|_{1} = \||\mathbb{E}_{0}(G^{(0)}(X_{\ell-i}))|_{q}\|_{1}$$

$$\leq \bar{\nu}\Big(\sup_{g \in L_{1/q,\tilde{C}}} \left|P^{\ell-i}(g) - \bar{\nu}(g)\right|\Big) \leq C(\ell-i)^{-(1-\gamma)/\gamma}. \quad (4.17)$$

We control now the quantity $\sum_{i=1}^{k} \sum_{\ell=i}^{k} \|\mathbb{E}_{0}(|\mathbb{E}_{i}(G^{(0)}(X_{\ell}))|_{q})\|_{(1-\gamma)/\gamma}$ with the help of (4.16) and (4.17). With this aim, we first write the following decomposition:

$$\sum_{i=1}^{k} \sum_{\ell=i}^{k} \|\mathbb{E}_{0}(|\mathbb{E}_{i}(G^{(0)}(X_{\ell}))|_{q})\|_{(1-\gamma)/\gamma} \leq \sum_{i=1}^{k} \sum_{\ell=2i+1}^{k} \||\mathbb{E}_{i}(G^{(0)}(X_{\ell}))|_{q}\|_{(1-\gamma)/\gamma} + \sum_{i=1}^{k} \sum_{\ell=i}^{2i} \|\mathbb{E}_{0}(|\mathbb{E}_{i}(G^{(0)}(X_{\ell}))|_{q}) - \mathbb{E}|\mathbb{E}_{i}(G^{(0)}(X_{\ell}))|_{q}\|_{(1-\gamma)/\gamma} + \sum_{i=1}^{k} \sum_{\ell=i}^{2i} \||\mathbb{E}_{i}(G^{(0)}(X_{\ell}))|_{q}\|_{1-\gamma} + \sum_{i=1}^{k} \sum_{\ell=i}^{2i} \||\mathbb{E}_{i}(G^{(0)}(X_{\ell}))|_{q}\|_{1-\gamma} + \sum_{i=1}^{k} \sum_{\ell=i}^{2i} \|\mathbb{E}_{i}(G^{(0)}(X_{\ell}))\|_{q}\|_{1-\gamma} + \sum_{i=1}^{k} \sum_{\ell=i}^{2i} \||\mathbb{E}_{i}(G^{(0)}(X_{\ell}))\|_{q}\|_{1-\gamma} + \sum_{i=1}^{k} \sum_{\ell=i}^{2i} \|\mathbb{E}_{i}(G^{(0)}(X_{\ell}))\|_{q}\|_{1-\gamma} + \sum_{i=1}^{k} \sum_{\ell=i}^{2i} \|\mathbb{E}_{i}(G^{(0)}(X_{\ell}))\|_{q}\|_{1-\gamma} + \sum_{i=1}^{k} \sum_{\ell=i}^{2i} \|\mathbb{E}_{i}(G^{(0)}(X_{\ell}))\|_{q}\|_{1-\gamma} + \sum_{i=1}^{k} \sum_{\ell=i}^{2i} \|\mathbb{E}_{i}(G^{(0)}(X_{\ell}))\|_{q}\|_{1-\gamma} + \sum_{i=1}^{k} \sum_{\ell=i}^{2i} \|\mathbb{E}_{i}(G^{(0)}(X_{\ell}))\|_{q} + \sum_{i=1}^{k} \sum_{i=i}^{2i} \|\mathbb{E}_{i}(G^{(0)}(X_{\ell}))\|_{q} + \sum_{i=i}^{k} \sum_{i=i}^{$$

Next, since $(1 - \gamma)/\gamma > 1$ and for any $i, |G^{(0)}(X_i)|_q \leq 1$ almost surely, we get

$$\sum_{i=1}^{k} \sum_{\ell=i}^{k} \|\mathbb{E}_{0}(|\mathbb{E}_{i}(G^{(0)}(X_{\ell}))|_{q})\|_{(1-\gamma)/\gamma} \leq \sum_{i=1}^{k} \sum_{\ell=2i+1}^{k} \||\mathbb{E}_{i}(G^{(0)}(X_{\ell}))|_{q}\|_{1}^{\gamma/(1-\gamma)} + 2^{\frac{1-2\gamma}{1-\gamma}} \sum_{i=1}^{k} \sum_{\ell=i}^{2i} \|\mathbb{E}_{0}(|\mathbb{E}_{i}(G^{(0)}(X_{\ell}))|_{q}) - \mathbb{E}|\mathbb{E}_{i}(G^{(0)}(X_{\ell}))|_{q}\|_{1}^{\gamma/(1-\gamma)} + \sum_{i=1}^{k} \sum_{\ell=i}^{2i} \||\mathbb{E}_{i}(G^{(0)}(X_{\ell}))|_{q}\|_{1}.$$

Therefore, using (4.16) and (4.17), we derive that

$$\sum_{i=1}^{k} \sum_{\ell=i}^{k} \|\mathbb{E}_{0}(|\mathbb{E}_{i}(G^{(0)}(X_{\ell}))|_{q})\|_{(1-\gamma)/\gamma} \\ \ll \sum_{i=1}^{k} \sum_{\ell=2i+1}^{k} \frac{1}{\ell-i} + \sum_{i=1}^{k} \sum_{\ell=i}^{2i} \frac{1}{i} + k + \sum_{i=1}^{k} \sum_{\ell=i+1}^{2i} \frac{1}{(\ell-i)^{\frac{1-\gamma}{\gamma}}} \ll k.$$
(4.18)

So starting from (4.14) and taking into account (4.15), (4.18) and the fact that $\gamma/(1-\gamma) < 1$, we get

$$\mathbb{E}\Big(\max_{1\leq k\leq n}\Big|\sum_{i=1}^{k} (G(X_i) - \mathbb{E}(G(X_i)))\Big|_q^{\frac{2(1-\gamma)}{\gamma}}\Big) \ll n + n\Big(\sum_{k=1}^{n} \frac{k^{\delta}}{k^{1+\delta\gamma/(1-\gamma)}}\Big)^{(1-\gamma)/(\delta\gamma)} \ll n^{(1-\gamma)/\gamma},$$

which completes the proof of (4.13) and then of the theorem. \Diamond

Proof of Theorem 4.2. We keep the same notations as in the proof of Theorem 4.1.

We start by proving Item 1. By (4.12), it suffices to prove that there exists a positive constant C such that for any $n \ge 1$,

$$\mathbb{E}\left(\max_{1\leq k\leq n} \left|\sum_{i=1}^{k} (G(X_i) - \mathbb{E}(G(X_i)))\right|_q^{1/\gamma}\right) \leq C n \log n.$$

$$(4.19)$$

Assume first that $\gamma = 1/2$. Applying Inequality (2.3), taking into account the stationarity and the fact that $|G(X_1) - \mathbb{E}(G(X_1))|_q \leq 1$ almost surely, we derive

$$\mathbb{E}\Big(\max_{1\leq k\leq n}\Big|\sum_{i=1}^{k} (G(X_i) - \mathbb{E}(G(X_i)))\Big|_{q}^{1/\gamma}\Big) \ll n + n\sum_{k=1}^{n} \||\mathbb{E}_{0}(G^{(0)}(X_k))|_{q}\|_{1}.$$

Therefore, using (4.17), it follows that

$$\mathbb{E}\Big(\max_{1\leq k\leq n}\Big|\sum_{i=1}^{k}(G(X_i)-\mathbb{E}(G(X_i)))\Big|_q^{1/\gamma}\Big)\ll n+n\sum_{k=1}^{n}k^{-1}.$$

proving (4.19) in the case $\gamma = 1/2$. We turn now to the proof of (4.19) when $\gamma \in (1/2, 1)$. With this aim, we apply the moment inequality (with $p = 1/\gamma$) stated in Proposition 5.1. This leads to

$$\mathbb{E}\Big(\max_{1\leq k\leq n}\Big|\sum_{i=1}^{k} (G(X_i) - \mathbb{E}(G(X_i)))\Big|_{q}^{1/\gamma}\Big) \leq C_{\gamma}n\sum_{k=0}^{n-1} (k+1)^{(1-2\gamma)/\gamma} ||\mathbb{E}_{0}(G^{(0)}(X_k))|_{q}||_{1},$$

where C_{γ} is a positive constant depending only on γ . Therefore, for any $\gamma \in (1/2, 1)$ using (4.17), we get

$$\mathbb{E}\Big(\max_{1\leq k\leq n}\Big|\sum_{i=1}^{k} (G(X_i) - \mathbb{E}(G(X_i)))\Big|_q^{1/\gamma}\Big) \leq \tilde{C}_{\gamma}n\Big(1 + \sum_{k=1}^{n-1} k^{-1}\Big),$$

proving (4.19) in case $\gamma \in (1/2, 1)$. This ends the proof of Item 1.

We turn now to the proof of Item 2. By (4.12), it suffices to prove that, for $\gamma \in [1/2, 1)$ and $p > 1/\gamma$, there exists a positive constant C such that for any $n \ge 1$,

$$\mathbb{E}\Big(\max_{1\le k\le n}\Big|\sum_{i=1}^{k} (G(X_i) - \mathbb{E}(G(X_i)))\Big|_q^p\Big) \le C n^{p+(\gamma-1)/\gamma}.$$
(4.20)

We shall distinguish two cases: $(p \ge 2 \text{ and } p > 1/\gamma)$ or $p \in [1/\gamma, 2[$. We first consider the case where $p \ge 2$ and $p > 1/\gamma$. To prove (4.20), we shall apply Inequality (2.3). Taking into account the stationarity and the fact that $|G(X_1) - \mathbb{E}(G(X_1))|_q \le 1$ almost surely, we derive

$$\mathbb{E}\Big(\max_{1\le k\le n}\Big|\sum_{i=1}^{k} (G(X_i) - \mathbb{E}(G(X_i)))\Big|_{q}^{p}\Big) \ll n^{p/2}\Big(\sum_{k=0}^{n} \||\mathbb{E}_{0}(G^{(0)}(X_k))|_{q}\|_{1}^{2/p}\Big)^{p/2}$$

Next, using (4.17) and the fact that $2(1-\gamma)/(\gamma p) < 1$, Inequality (4.20) follows.

We consider now the case where $p \in [1/\gamma, 2[$. Using, once again, the moment inequality stated in Proposition 5.1, we get

$$\mathbb{E}\Big(\max_{1\leq k\leq n}\Big|\sum_{i=1}^{k} (G(X_i) - \mathbb{E}(G(X_i)))\Big|_{q}^{p}\Big) \leq C_{p}n\sum_{k=0}^{n-1} (k+1)^{p-2} \||\mathbb{E}_{0}(G^{(0)}(X_k))|_{q}\|_{1},$$

where C_p is a positive constant depending only on p. Using then (4.17) and the fact that $p > 1/\gamma$, (4.20) follows. This ends the proof of the theorem. \diamond

Proof of Theorem 4.3. We keep the same notations as in the proof of Theorem 4.1. Notice first that, for any non-negative x,

$$\nu\Big(\max_{1\leq k\leq n} D_{k,q} \geq x\Big) = \bar{\nu}\Big(\max_{1\leq k\leq n} \Big|\int_0^1 \Big|\sum_{i=1}^k (f_t \circ T^i \circ \pi - \bar{\nu}(f_t \circ \pi)\Big|^q dt\Big|^{1/q} \geq x\Big)$$
$$= \bar{\nu}\Big(\max_{1\leq k\leq n} \Big|\int_0^1 \Big|\sum_{i=1}^k (f_t \circ \pi \circ \bar{T}^i - \bar{\nu}(f_t \circ \pi)\Big|^q dt\Big|^{1/q} \geq x\Big)$$
$$= \bar{\nu}\Big(\max_{1\leq k\leq n} \Big|\sum_{i=1}^k (G(\bar{T}^i) - \bar{\nu}(G(\bar{T}^i)))\Big|_q \geq x\Big).$$

According to (4.11),

$$\nu\Big(\max_{1\leq k\leq n} D_{k,q} \geq x\Big) = \mathbb{P}\Big(\max_{1\leq k\leq n} \Big|\sum_{i=k}^{n} (G(X_i) - \mathbb{E}(G(X_i)))\Big|_q \geq x\Big)$$
$$\leq \mathbb{P}\Big(\max_{1\leq k\leq n} \Big|\sum_{i=1}^{k} (G(X_i) - \mathbb{E}(G(X_i)))\Big|_q \geq x/2\Big).$$

The theorem will then follow if we can prove that, for any positive real x,

$$\mathbb{P}\Big(\max_{1\le k\le n} \Big|\sum_{i=1}^{k} (G(X_i) - \mathbb{E}(G(X_i)))\Big|_q \ge 4x\Big) \ll nx^{-1/\gamma}.$$
(4.21)

Obviously, it suffices to prove it for $x \le n/4$ since otherwise the left-hand side is zero so the inequality is trivial. To prove (4.21) when $x \le n/4$, we apply Inequality (5.4) with $q = \max([x], 1)$. Using (4.17), this leads to the following inequality: for any positive real $x \le n/4$,

$$\mathbb{P}\Big(\max_{1\le k\le n}\Big|\sum_{i=1}^{k} (G(X_i) - \mathbb{E}(G(X_i)))\Big|_q \ge 4x\Big) \ll \frac{n}{x^{1/\gamma}} + \frac{n}{x^2} \sum_{k=0}^{|x|} \frac{1}{(k+1)^{(1-\gamma)/\gamma}},$$

and (4.21) follows.

Proof of Theorem 4.5. We start by proving that Item 1 of Theorem 4.4 holds for $S_n(f) = \sum_{i=1}^n (f \circ T_{\gamma}^i - \nu_{\gamma}(f))$ where T_{γ} is a GPM map of parameter $\gamma \in (1/2, 1)$ and invariant measure ν_{γ} , and f is a BV function. With this aim, we first note that when $(X_i)_{i \in \mathbb{Z}}$ is a stationary sequence of real-valued random variables adapted to an increasing filtration (\mathcal{F}_i) and such that $\mathbb{P}(|X_0| \leq M) = 1$, Theorem 6 and Lemma 19 in [15] imply the following moment inequality for the maximum of the partial sums of $(X_i)_{i \in \mathbb{Z}}$: for any p > 2 and any positive integer n,

$$\mathbb{E}\Big(\max_{1\leq k\leq n} \big|\sum_{i=1}^{k} X_{i}\big|^{p}\Big) \ll n^{p/2} \Big(\sum_{k=0}^{n-1} \lambda_{k}\Big)^{p/2} + n \|X_{0}\|_{p}^{p} + n\Big(\sum_{k=1}^{n} \frac{1}{k^{1/p}} \lambda_{k}^{1/p}\Big)^{p} + n\Big(\sum_{k=1}^{n} \frac{1}{k^{1+(2\delta)/p}} \Big(\sum_{j=1}^{k} j\lambda_{j}^{2/p}\Big)^{\delta}\Big)^{p/(2\delta)}, \quad (4.22)$$

where $\delta = \min(1, 1/(p-2))$ and

$$\lambda_j = \max\left(\|\mathbb{E}_0(X_j)\|_1, \sup_{i \ge j} \|\mathbb{E}_0(X_i X_j) - \mathbb{E}(X_i X_j)\|_1\right), \text{ for } j \ge 0.$$
(4.23)

The constant that is implicitly involved in the notation \ll in (4.22) depends only on p, δ and M. Let now $(Y_i)_{i\in\mathbb{N}}$ be a stationary Markox chain defined on a probability space $(\Omega, \mathcal{A}, \mathbb{P})$ with state space [0, 1], invariant measure ν_{γ} and transition kernel K_{γ} given by the Perron-Frobenius operator of T_{γ} with respect to ν_{γ} . Let $X_k = f(Y_k) - \nu_{\gamma}(f)$. Since the law of $(T_{\gamma}, \ldots, T_{\gamma}^n)$ under ν_{γ} is the same as that of (Y_n, \ldots, Y_1) under \mathbb{P} , it follows that for any positive real x,

$$\nu_{\gamma}\Big(\max_{1\le k\le n} |S_k(f)| \ge x\Big) \le \mathbb{P}\Big(2\max_{1\le k\le n} \Big|\sum_{i=1}^k X_i\Big| \ge x\Big), \tag{4.24}$$

implying that, for any real $r \in [1, \infty[$,

$$\left\| \max_{1 \le k \le n} |S_k(f)| \right\|_{r,\nu_{\gamma}}^r \le 2^r \mathbb{E} \left(\max_{1 \le k \le n} \left| \sum_{i=1}^k X_i \right|^r \right).$$

$$(4.25)$$

Starting from (4.25) with r = p, using (4.22) and taking into account that there exists a positive constant C such that for any $k \ge 1$,

$$\lambda_k \le \frac{C}{k^{(1-\gamma)/\gamma}} \tag{4.26}$$

(see Proposition 1.17 in [4]), Item 1 easily follows. To prove Item 2 (resp. Item 3) it suffices to start from (4.25) (resp. (4.24)) and to use Inequality (5.5) when $p \in]1/\gamma, 2[$ or Inequality (2.3) when $p \geq 2$ and is strictly larger than $1/\gamma$ (resp. Inequality (5.4)) and to use the upper bound (4.26). \diamond

5 Appendix

5.1 A Rosenthal-type inequality for stationary sequences

In this section, for the reader convenience, we recall the Rosenthal-type inequality stated in [6] (see Inequality (3.11) therein). This inequality is the extension to Banach-valued random variables of the Rosenthal type inequality given by Merlevède and Peligrad [15].

Let $(\Omega, \mathcal{A}, \mathbb{P})$ be a probability space, and $\theta : \Omega \mapsto \Omega$ be a bijective bimeasurable transformation preserving the probability \mathbb{P} . For a σ -algebra \mathcal{F}_0 satisfying $\mathcal{F}_0 \subseteq T^{-1}(\mathcal{F}_0)$, we define the nondecreasing filtration $(\mathcal{F}_i)_{i \in \mathbb{Z}}$ by $\mathcal{F}_i = \theta^{-i}(\mathcal{F}_0)$. We shall use the notations $\mathbb{E}_k(\cdot) = \mathbb{E}(\cdot | \mathcal{F}_k)$.

Let X_0 be a random variable with values in \mathbb{B} . Define the stationary sequence $(X_i)_{i \in \mathbb{Z}}$ by $X_i = X_0 \circ T^i$, and the partial sum S_n by $S_n = X_1 + X_2 + \cdots + X_n$.

Theorem 5.1. Assume that X_0 belongs to $\mathbb{L}^p(\mathbb{B})$ where $(\mathbb{B}, |\cdot|_{\mathbb{B}})$ is a separable Banach space and p is a real number in $]2, \infty[$. Assume that X_0 is \mathcal{F}_0 -measurable. Then, for any $r \ge 0$,

$$\mathbb{E}\Big(\max_{1 \le j \le 2^r} |S_j|_{\mathbb{B}}^p\Big) \ll 2^r \mathbb{E}(|X_0|_{\mathbb{B}}^p) + 2^r \left(\sum_{k=0}^{r-1} \frac{\|\mathbb{E}_0(|S_{2^k}|_{\mathbb{B}}^2)\|_{p/2}^{\delta}}{2^{2\delta k/p}}\right)^{p/(2\delta)},$$
(5.1)

where $\delta = \min(1/2, 1/(p-2)).$

Remark 5.1. The inequality in the above theorem implies that for any positive integer n,

$$\mathbb{E}\left(\max_{1\leq j\leq n}|S_j|_{\mathbb{B}}^p\right) \ll n\mathbb{E}(|X_0|_{\mathbb{B}}^p) + n\left(\sum_{k=1}^n \frac{1}{k^{1+2\delta/p}} \|\mathbb{E}_0(|S_k|_{\mathbb{B}}^2)\|_{p/2}^{\delta}\right)^{p/(2\delta)}.$$
(5.2)

5.2 A deviation inequality

The following proposition is adapted from Proposition 4 in [5]. It also extends Proposition 6.1 in [2] to random variables taking values in a separable Banach space belonging to the class $C_2(2, c_2)$.

Proposition 5.1. Let Y_1, Y_2, \ldots, Y_n be *n* random variables with values in a separable Banach space $(\mathbb{B}, |\cdot|_{\mathbb{B}})$ belonging to the class $C_2(2, c_2)$. Assume that $\mathbb{P}(|Y_k|_{\mathbb{B}} \leq M) = 1$ for any $k \in \{1, \ldots, n\}$. Let $\mathcal{F}_1, \ldots, \mathcal{F}_n$ be an increasing filtration such that Y_k is \mathcal{F}_k -measurable for any $k \in \{1, \ldots, n\}$. Let $S_n = \sum_{k=1}^n Y_k$, and for $k \in \{0, \ldots, n-1\}$, let

$$\theta(k) = \max\left\{ \mathbb{E}(|\mathbb{E}(Y_i|\mathcal{F}_{i-k})|_{\mathbb{B}}), i \in \{k+1,\dots,n\} \right\}.$$
(5.3)

Then, for any $q \in \{1, ..., n\}$, and any $x \ge qM$, the following inequality holds

$$\mathbb{P}\Big(\max_{1\leq k\leq n}|S_k|_{\mathbb{B}}\geq 4x\Big)\leq \frac{n\theta(q)}{x}\mathbf{1}_{q< n}+\frac{4c_2K^2nM}{x^2}\sum_{k=0}^{q-1}\theta(k)\,,\tag{5.4}$$

where $K = \sqrt{\max(c_2, 1)}$. In addition, for any $p \in [1, 2[$,

$$\mathbb{E}\Big(\max_{1\le k\le n} |S_k|_{\mathbb{B}}^p\Big) \le \Big(4^p p + \frac{4^{p+1}pc_2K^2}{2-p}\Big)M^{p-1}n\sum_{k=0}^{n-1}(k+1)^{p-2}\theta(k)\,.$$
(5.5)

Proof of Proposition 5.1. Let $S_0 = 0$ and define the random variables U_i by: $U_i = S_{iq} - S_{(i-1)q}$ for $i \in \{1, \ldots, \lfloor n/q \rfloor\}$ and $U_{\lfloor n/q \rfloor+1} = S_n - S_{q \lfloor n/q \rfloor}$. By Proposition 4 in [5], for any $x \ge Mq$,

$$\mathbb{P}\left(\max_{1\leq k\leq n}|S_{k}|_{\mathbb{B}}\geq 4x\right)\leq \frac{1}{x}\sum_{i=3}^{[n/q]+1}\mathbb{E}(|\mathbb{E}(U_{i}|\mathcal{F}_{(i-2)q})|_{\mathbb{B}})+\frac{c_{2}}{x^{2}}\sum_{i=1}^{[n/q]+1}\mathbb{E}(|U_{i}-\mathbb{E}(U_{i}|\mathcal{F}_{(i-2)q})|_{\mathbb{B}})\\ \leq \frac{1}{x}\sum_{i=3}^{[n/q]+1}\mathbb{E}(|\mathbb{E}(U_{i}|\mathcal{F}_{(i-2)q})|_{\mathbb{B}})+\frac{4c_{2}}{x^{2}}\sum_{i=1}^{[n/q]+1}\mathbb{E}(|U_{i}|_{\mathbb{B}}^{2}).$$
(5.6)

Since $(\theta(k))_{k\geq 0}$ is a non-increasing sequence, it is not hard to see that

$$\sum_{i=3}^{[n/q]+1} \mathbb{E}(|\mathbb{E}(U_i|\mathcal{F}_{(i-2)q})|_{\mathbb{B}}) \le n\theta(q)\mathbf{1}_{q < n}.$$
(5.7)

To handle the second term in (5.6), we use Inequality (2.2) with p = 2. This leads to the following upper bounds: for any $i \in \{1, \ldots, \lfloor n/q \rfloor\}$,

$$\mathbb{E}(|U_i|_{\mathbb{B}}^2) \le K^2 \sum_{k=(i-1)q+1}^{iq} \sum_{j=k}^{iq} \mathbb{E}(|Y_k|_{\mathbb{B}}|\mathbb{E}(Y_j|\mathcal{F}_k)|_{\mathbb{B}}),$$

and

$$\mathbb{E}(|U_{[n/q]+1}|_{\mathbb{B}}^2) \le K^2 \sum_{k=q[n/q]+1}^n \sum_{j=k}^n \mathbb{E}(|Y_k|_{\mathbb{B}}|\mathbb{E}(Y_j|\mathcal{F}_k)|_{\mathbb{B}}),$$

where $K = \sqrt{\max(c_2, 1)}$. Using the fact that $\mathbb{P}(|Y_k|_{\mathbb{B}} \leq M) = 1$ for any $k \in \{1, \ldots, n\}$ and that $(\theta(k))_{k\geq 0}$ is a non-increasing sequence, we then derive that, for any $i \in \{1, \ldots, [n/q]\}$,

$$\mathbb{E}(|U_i|_{\mathbb{B}}^2) \le K^2 M \sum_{k=(i-1)q+1}^{iq} \sum_{j=k}^{iq} \theta(j-k) \le K^2 M q \sum_{k=0}^{q-1} \theta(k) ,$$

and

$$\mathbb{E}(|U_{[n/q]+1}|_{\mathbb{B}}^2) \le K^2 M \sum_{k=q[n/q]+1}^n \sum_{j=k}^n \theta(j-k) \le K^2 M(n-q[n/q]) \sum_{k=0}^{q-1} \theta(k) \,.$$

Whence

$$\sum_{i=1}^{[n/q]+1} \mathbb{E}(|U_i|_{\mathbb{B}}^2) \le K^2 M n \sum_{k=0}^{q-1} \theta(k) \,.$$
(5.8)

Starting from (5.6) and using the upper bounds (5.7) and (5.8), Proposition 5.1 follows. \Diamond

5.3 A maximal inequality

Proposition 5.2. Let $n \ge 2$ be an integer and Y_1, Y_2, \ldots, Y_n be n random variables with values in a separable Banach space $(\mathbb{B}, |\cdot|_{\mathbb{B}})$. Assume that $\mathbb{P}(|Y_k|_{\mathbb{B}} \le M) = 1$ for any $k \in \{1, \ldots, n\}$. Let $\mathcal{F}_1, \ldots, \mathcal{F}_n$ be an increasing filtration such that Y_k is \mathcal{F}_k -measurable for any $k \in \{1, \ldots, n\}$. Let $S_n = \sum_{k=1}^n Y_k$ and $\theta(k)$ be defined by (5.3). Then, for any real p > 1, the following inequality holds:

$$\mathbb{E}\Big(\max_{1\le k\le n} |S_k|_{\mathbb{B}}^p\Big) \le \frac{1}{2} \Big(\frac{2p}{p-1}\Big)^p \mathbb{E}(|S_n|_{\mathbb{B}}^p) + 2^{p-1} 3^p p M^{p-1} n \sum_{k=0}^{n-2} (k+1)^{p-2} \theta(k) \,.$$

Proof of Proposition 5.2. All along the proof, $\mathbb{E}_k(\cdot) = \mathbb{E}(\cdot|\mathcal{F}_k)$. We start by noticing that

$$S_k = \mathbb{E}_k(S_n) + \mathbb{E}_k(S_k - S_n) \,.$$

Therefore

$$\mathbb{E}\Big(\max_{1\leq k\leq n}|S_k|_{\mathbb{B}}^p\Big)\leq 2^{p-1}\mathbb{E}\Big(\max_{1\leq k\leq n}|\mathbb{E}_k(S_n)|_{\mathbb{B}}^p\Big)+2^{p-1}\mathbb{E}\Big(\max_{1\leq k\leq n}|\mathbb{E}_k(S_n-S_k)|_{\mathbb{B}}^p\Big).$$

Notice now that $(|\mathbb{E}_k(S_n)|, \mathcal{F}_k)_{1 \le k \le n}$ is a submartingale. Therefore by the Doob's maximal inequality,

$$\mathbb{E}\Big(\max_{1\leq k\leq n} |\mathbb{E}_k(S_n)|_{\mathbb{B}}^p\Big) \leq \Big(\frac{p}{p-1}\Big)^p \mathbb{E}(|S_n|_{\mathbb{B}}^p).$$

So, overall,

$$\mathbb{E}\Big(\max_{1\leq k\leq n}|S_k|_{\mathbb{B}}^p\Big)\leq 2^{-1}\Big(\frac{2p}{p-1}\Big)^p\mathbb{E}(|S_n|_{\mathbb{B}}^p)+2^{p-1}\mathbb{E}\Big(\max_{1\leq k\leq n}|\mathbb{E}_k(S_n-S_k)|_{\mathbb{B}}^p\Big).$$

To end the proposition, it remains to prove that

$$\mathbb{E}\Big(\max_{1\le k\le n} |\mathbb{E}_k(S_n - S_k)|_{\mathbb{B}}^p\Big) \le 3^p p M^{p-1} n \sum_{k=0}^{n-2} (k+1)^{p-2} \theta(k) \,.$$
(5.9)

With this aim, we write

$$\mathbb{E}\Big(\max_{1\leq k\leq n} |\mathbb{E}_k(S_n - S_k)|_{\mathbb{B}}^p\Big) = p \int_0^{nM} x^{p-1} \mathbb{P}\Big(\max_{1\leq k\leq n} |\mathbb{E}_k(S_n - S_k)|_{\mathbb{B}} > x\Big) dx.$$

Let q be a non-negative integer such that $q \leq n$. Notice that

$$|\mathbb{E}_k(S_n - S_k)|_{\mathbb{B}} = \Big|\sum_{i=k+1}^n \mathbb{E}_k(X_i)\Big|_{\mathbb{B}} \le \Big|\sum_{i=k+1}^n \mathbb{E}_k(X_i - \mathbb{E}_{i-q}(X_i))\Big|_{\mathbb{B}} + \Big|\sum_{i=k+1}^n \mathbb{E}_k(\mathbb{E}_{i-q}(X_i))\Big|_{\mathbb{B}}.$$

But

$$\Big|\sum_{i=k+1}^{n} \mathbb{E}_k(X_i - \mathbb{E}_{i-q}(X_i))\Big|_{\mathbb{B}} = \Big|\sum_{i=k+1}^{q+k} (\mathbb{E}_k(X_i) - \mathbb{E}_{i-q}(X_i))\Big|_{\mathbb{B}} \le 2qM$$

Therefore, for any real x such that $x \in [0, n]$, choosing q = [x], we get

$$\mathbb{P}\Big(\max_{1\leq k\leq n} |\mathbb{E}_k(S_n - S_k)|_{\mathbb{B}} > 3Mx\Big) \leq \mathbb{P}\Big(\max_{1\leq k\leq n} \Big|\sum_{i=k+1}^n \mathbb{E}_k(\mathbb{E}_{i-[x]}(X_i))\Big|_{\mathbb{B}} > Mx\Big)$$
$$\leq \mathbb{P}\Big(\max_{1\leq k\leq n} \mathbb{E}_k\Big(\sum_{i=2}^n |\mathbb{E}_{i-[x]}(X_i)|_{\mathbb{B}}\Big) > Mx\Big).$$

But $(\mathbb{E}_k(\sum_{i=2}^n |\mathbb{E}_{i-[x]}(X_i)|), \mathcal{F}_k)_{1 \le k \le n}$ is a martingale, so the Doob-Kolmogorov's inequality implies

$$\mathbb{P}\Big(\max_{1\leq k\leq n} \mathbb{E}_k\Big(\sum_{i=2}^n |\mathbb{E}_{i-[x]}(X_i)|_{\mathbb{B}}\Big) > Mx\Big) \leq \frac{1}{Mx} \sum_{i=2}^n \mathbb{E}\Big(|\mathbb{E}_{i-[x]}(X_i)|_{\mathbb{B}}\Big) \leq \frac{n\theta([x])}{Mx}.$$

So, overall,

$$\mathbb{E}\Big(\max_{1 \le k \le n} |\mathbb{E}_k (S_n - S_k)|_{\mathbb{B}}^p\Big) = p(3M)^p \int_0^{n/3} x^{p-1} \mathbb{P}\Big(\max_{1 \le k \le n} |\mathbb{E}_k (S_n - S_k)|_{\mathbb{B}} > 3Mx\Big) dx$$
$$\leq 3^p p M^{p-1} n \int_0^{n/3} x^{p-2} \theta([x]) dx \,,$$

proving (5.9) by using the fact that $(\theta(k))_k$ is a non-increasing sequence. The proof of the proposition is therefore complete. \diamond

5.4 Proof of Inequality (2.3)

Proposition 5.2 together with Inequality (2.2) leads to

$$\mathbb{E}(\max_{1 \le k \le n} |S_k|_{\mathbb{B}}^p) \le 2^{-1} \left(\frac{2p}{p-1}\right)^p K^p \left(\sum_{i=1}^n \max_{i \le \ell \le n} \left\| |X_i|_{\mathbb{B}} \right| \sum_{k=i}^\ell \mathbb{E}(X_k|\mathcal{F}_i) \Big|_{\mathbb{B}} \right\|_{p/2} \right)^{p/2} + 2^{p-1} 3^p p M^{p-1} n \sum_{k=0}^{n-2} (k+1)^{p-2} \theta(k). \quad (5.10)$$

Since $\mathbb{P}(|X_k|_{\mathbb{B}} \leq M) = 1$ for any $k \in \{1, \ldots, n\}$, it follows that

$$\sum_{i=1}^{n} \max_{i \le \ell \le n} \left\| |X_i|_{\mathbb{B}} \right| \sum_{k=i}^{\ell} \mathbb{E}(X_k | \mathcal{F}_i) \Big|_{\mathbb{B}} \right\|_{p/2} \le n M^{2-2/p} \sum_{k=0}^{n-1} \theta^{2/p}(k) \,. \tag{5.11}$$

On the other hand, since $(\theta(k))_{k\geq 1}$ is non-increasing,

$$\sum_{k=1}^{n-2} (k+1)^{p-2} \theta(k) = \sum_{\ell=0}^{\log_2(n-1)-1} \sum_{k=2^{\ell}}^{2^{\ell+1}-1} (k+1)^{p-2} \theta(k) \le 2^{p-2} \sum_{\ell=0}^{\log_2(n-1)} 2^{\ell(p-1)} \theta(2^{\ell}).$$

Hence, using the fact that $p \geq 2$ and again that $(\theta(k))_{k\geq 1}$ is non-increasing, we successively derive

$$\sum_{k=1}^{n-2} (k+1)^{p-2} \theta(k) \le 2^{p-2} \Big(\sum_{\ell=0}^{\log_2(n-1)} 2^{\ell(2-2/p)} \theta^{2/p} (2^\ell) \Big)^{p/2} \le 2^{p-2} \Big(\theta^{2/p}(1) + 2 \sum_{\ell=1}^{\log_2(n-1)} \sum_{k=2^{\ell-1}+1}^{2^\ell} 2^{\ell(1-2/p)} \theta^{2/p} (2^\ell) \Big)^{p/2} \le 2^{2p-3} \Big(\sum_{k=1}^{n-1} k^{1-2/p} \theta^{2/p}(k) \Big)^{p/2}.$$

Since $p \ge 2$, it follows that

$$\sum_{k=1}^{n-2} (k+1)^{p-2} \theta(k) \le 2^{2p-3} n^{p/2-1} \Big(\sum_{k=1}^{n-1} k^{1-2/p} \theta^{2/p}(k) \Big)^{p/2}.$$
(5.12)

Starting from (5.10) and considering the upper bounds (5.11) and (5.12), the inequality (2.3) follows. \diamond

5.5 Dependence properties of Young towers

In this section, we assume that T is a nonuniformly expanding map on (\mathcal{X}, λ) with λ a probability measure on \mathcal{X} , and that T can be modelled by a Young tower. As in Section 4.4, \mathcal{X} can be any bounded metric space and not necessarily the unit interval.

Proposition 5.3. Let T be map that can be modelled by a Young tower with polynomial tails of the return times of order $1/\gamma$ with $\gamma \in (0,1)$. Then the inequality (4.3) holds, that is: for any $\alpha \in (0,1]$ there exists $K_{\alpha} > 0$ such that

$$\bar{\nu}\Big(\sup_{f\in L_{\alpha,1}}|P^n(f)-\bar{\nu}(f)|\Big)\leq \frac{K_{\alpha}}{n^{(1-\gamma)/\gamma}}\,.$$

Proof of Proposition 5.3. The proof is a slight modification of the proof of Theorem 2.3.6 in [8] and is included here for the sake of completeness. In this proof, C is a positive constant, and C_{α} is a positive constant depending only on α . Both constants may vary from line to line.

We keep the same notations as in Subsection 4.1. For $f \in L_{\alpha}$, let

$$\|f\|_{L_{\alpha}} = L_{\alpha}(f) + \|f\|_{\infty}$$

Let $f^{(0)} = f - \bar{\nu}(f)$. Since $||f^{(0)}||_{\infty} \leq L_{\alpha}(f)$, it follows that

$$\|f - \bar{\nu}(f)\|_{L_{\alpha}} \le 2L_{\alpha}(f)$$
. (5.13)

Recall that one has the decomposition

$$P^{n}f = \sum_{a+k+b=n} \lambda_{b}(f)A_{a}(\mathbf{1}_{\bar{Y}}) + \sum_{a+k+b=n} A_{a}E_{k}B_{b}f + C_{n}f, \qquad (5.14)$$

where the operators A_n , B_n , C_n and E_n and are defined in Chapter 2 of Gouëzel's PhD thesis [8] and $\lambda_b(f) = \bar{\nu}(B_b(f))$. In particular, Gouëzel has proved that

$$\|E_k f\|_{L_{\alpha}} \le \frac{C_{\alpha} \|f\|_{L_{\alpha}}}{(k+1)^{(1-\gamma)/\gamma}} \quad \text{and} \quad \|B_k f\|_{L_{\alpha}} \le \frac{C_{\alpha} \|f\|_{L_{\alpha}}}{(k+1)^{1/\gamma}}.$$
(5.15)

Following the proof of Lemma 2.3.5 in [8], there exists a set Z_n such that, for any bounded measurable function g,

$$|C_n(g)| \le C ||g||_{\infty} \mathbf{1}_{Z_n} \,, \tag{5.16}$$

and

$$\bar{\nu}(Z_n) \le \frac{C}{(n+1)^{(1-\gamma)/\gamma}}$$
 (5.17)

We now turn to the term $\sum_{a+k+b=n} A_a E_k B_b f$ in (5.14). Following the proof of Lemma 2.3.3. in [8], there exist a set U_n such that, for any bounded measurable function g,

$$|A_n(g)| \le C ||g||_{\infty} \mathbf{1}_{U_n} \,, \tag{5.18}$$

and

$$\bar{\nu}(U_n) \le \frac{C}{(n+1)^{1/\gamma}}$$
 (5.19)

Using successively (5.18) and (5.15), we obtain that

$$\left|\sum_{a+k+b=n} A_{a} E_{k} B_{b} f\right| \leq C \sum_{a+k+b=n} \|E_{k} B_{b} f\|_{\infty} \mathbf{1}_{U_{a}}$$
$$\leq C_{\alpha} \sum_{a+k+b=n} \|B_{b} f\|_{L_{\alpha}} \frac{\mathbf{1}_{U_{a}}}{(k+1)^{(1-\gamma)/\gamma}}$$
$$\leq C_{\alpha} \|f\|_{L_{\alpha}} \sum_{a+k+b=n} \frac{\mathbf{1}_{U_{a}}}{(k+1)^{(1-\gamma)/\gamma}(b+1)^{1/\gamma}}.$$
(5.20)

We now turn to the term $\sum_{a+k+b=n} A_a(\mathbf{1}_{\bar{Y}}) \cdot \bar{\nu}(B_b f)$ in (5.14). From the last equality of (2.21) in [8], if $\bar{\nu}(f) = 0$,

$$\left|\sum_{b=0}^{n-a} \bar{\nu}(B_b f)\right| = \left|\sum_{b>n-a} \bar{\nu}(B_b f)\right| \le \sum_{b>n-a} \|B_b f\|_{L_{\alpha}} \le \sum_{b>n-a} \frac{C_{\alpha} \|f\|_{L_{\alpha}}}{(b+1)^{1/\gamma}} \le \frac{C_{\alpha} \|f\|_{L_{\alpha}}}{(n+1-a)^{(1-\gamma)/\gamma}}.$$
(5.21)

From (5.21) and (5.18), if $\bar{\nu}(f) = 0$,

$$\left|\sum_{a=0}^{n} A_{a}(\mathbf{1}_{\bar{Y}}) \cdot \left(\sum_{b=0}^{n-a} \bar{\nu}(B_{b}f)\right)\right| \leq C_{\alpha} \|f\|_{L_{\alpha}} \sum_{a=0}^{n} \frac{\mathbf{1}_{U_{a}}}{(n+1-a)^{(1-\gamma)/\gamma}}.$$
(5.22)

From (5.13), $||f - \nu(f)||_{L_{\alpha}} \leq 2L_{\alpha}(f)$. Hence, it follows from (5.14), (5.16), (5.20) and (5.22) that

$$|P^{n}(f-\nu(f))| \leq C_{\alpha}L_{\alpha}(f) \Big(\mathbf{1}_{Z_{n}} + \sum_{a=0}^{n} \frac{\mathbf{1}_{U_{a}}}{(n+1-a)^{(1-\gamma)/\gamma}} + \sum_{a+k+b=n} \frac{\mathbf{1}_{U_{a}}}{(k+1)^{(1-\gamma)/\gamma}(b+1)^{1/\gamma}} \Big).$$
(5.23)

From (5.23), (5.17) and (5.19), it follows that

$$\bar{\nu}\Big(\sup_{f\in L_{\alpha,1}}|P^{n}(f)-\bar{\nu}(f)|\Big) \leq C_{\alpha}\Big(\frac{1}{(n+1)^{(1-\gamma)/\gamma}} + \sum_{a=0}^{n}\frac{1}{(a+1)^{1/\gamma}(n+1-a)^{(1-\gamma)/\gamma}} + \sum_{a+k+b=n}\frac{1}{(a+1)^{1/\gamma}(k+1)^{(1-\gamma)/\gamma}(b+1)^{1/\gamma}}\Big).$$
 (5.24)

All the sums on right hand being of the same order (see the end of the proof of Proposition 6.2 in [2]), it follows that there exists $K_{\alpha} > 0$ such that

$$\bar{\nu}\Big(\sup_{f\in L_{\alpha,1}}|P^n(f)-\bar{\nu}(f)|\Big)\leq \frac{K_\alpha}{n^{(1-\gamma)/\gamma}}\,,$$

and the proof is complete. \Diamond

5.6 Proof of Lemma 1.1

We shall only prove Item 1 since the proof of Item 2 uses the same arguments as for the \mathbb{L}^2 case. Set $|x|_q = \left(\int_{\mathcal{X}} |x(t)|^q d\nu(t)\right)^{1/q}$ and recall that $\psi_p(x) = |x|_q^p$. Proceeding as in the proof of Proposition 2.1 in Pinelis (1994), we infer that, for any x and u in \mathbb{L}^q ,

$$D\psi_2(x)(u) = 2(\psi_2(x))^{1-q/2} \int_{\mathcal{X}} u(t)|x(t)|^{q-2}x(t)\mu(dt)$$

and

$$D^{2}\psi_{2}(x)(u,u) = 2(q-1)(\psi_{2}(x))^{1-q/2} \int_{\mathcal{X}} u^{2}(t)|x(t)|^{q-2}\mu(dt) + 2(2-q)(\psi_{2}(x))^{1-q} \left(\int_{\mathcal{X}} u(t)x(t)|x(t)|^{q-2}\mu(dt)\right)^{2}.$$

Hence by the chain rule, it follows that

$$D^{2}\psi_{p}(x)(u,u) = p(q-1)|x|_{q}^{p-q} \int_{\mathcal{X}} u^{2}(t)|x(t)|^{q-2}\mu(dt) + p(p-q)|x|_{q}^{p-2q} \left(\int_{\mathcal{X}} u(t)x(t)|x(t)|^{q-2}\mu(dt)\right)^{2}, \quad (5.25)$$

Item 1 then follows by using Hölder's inequality (and, for the case q > p, by taking into account the fact that $\left(\int_{\mathcal{X}} v(t)|x(t)|^{q-2}x(t)\mu(dt)\right)^2$ is non-negative). \diamond

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