

# SOME UNBOUNDED FUNCTIONS OF INTERMITTENT MAPS FOR WHICH THE CENTRAL LIMIT THEOREM HOLDS

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**Abstract.** We compute some dependence coefficients for the stationary Markov chain whose transition kernel is the Perron-Frobenius operator of an expanding map  $T$  of  $[0, 1]$  with a neutral fixed point. We use these coefficients to prove a central limit theorem for the partial sums of  $f \circ T^i$ , when  $f$  belongs to a large class of unbounded functions from  $[0, 1]$  to  $\mathbb{R}$ . We also prove other limit theorems and moment inequalities.

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## 1. INTRODUCTION

For  $\gamma$  in  $]0, 1[$ , we consider the intermittent map  $T_\gamma$  from  $[0, 1]$  to  $[0, 1]$ , studied for instance by Liverani, Saussol and Vaienti (1999), which is a modification of the Pomeau-Manneville map (1980):

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

We denote by  $\nu_\gamma$  the unique  $T_\gamma$ -probability measure on  $[0, 1]$ . We denote by  $K_\gamma$  the Perron-Frobenius operator of  $T_\gamma$  with respect to  $\nu_\gamma$ : for any bounded measurable functions  $f, g$ ,

$$\nu_\gamma(f \cdot g \circ T_\gamma) = \nu_\gamma(K_\gamma(f)g).$$

Let  $(X_i)_{i \geq 0}$  be a stationary Markov chain with invariant measure  $\nu_\gamma$  and transition Kernel  $K_\gamma$ . It is well known (see for instance Lemma XI.3 in Hennion and Hervé (2001)) that on the probability space  $([0, 1], \nu_\gamma)$ , the random variable  $(T_\gamma, T_\gamma^2, \dots, T_\gamma^n)$  is distributed as  $(X_n, X_{n-1}, \dots, X_1)$ . Hence any information on the law of

$$S_n(f) = \sum_{i=1}^n f \circ T_\gamma^i$$

can be obtained by studying the law of  $\sum_{i=1}^n f(X_i)$ .

In 1999, Young proved that such systems (among many others) may be described by a Young tower with polynomial decay of the return time. From this construction, she was able to control the covariances  $\nu_\gamma(f \circ T^n \cdot (g - \nu_\gamma(g)))$  for any bounded function  $f$  and any  $\alpha$ -Hölder function  $g$ , and then to prove that  $n^{-1/2}(S_n(f) - \nu_\gamma(f))$  converges in distribution to a normal law as soon as  $\gamma < 1/2$  and  $f$  is any  $\alpha$ -Hölder function. For  $\gamma = 1/2$ , Gouëzel (2004) proved that the central limit theorem remains true with the same normalization  $\sqrt{n}$  if  $f(0) = \nu_\gamma(f)$ , and with the normalization

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$\sqrt{n \ln(n)}$  if  $f(0) \neq \nu_\gamma(f)$ . When  $1/2 < \gamma < 1$ , he proved that if  $f$  is  $\alpha$ -Hölder and  $f(0) \neq \nu_\gamma(f)$ ,  $n^{-\gamma}(S_n(f) - \nu_\gamma(f))$  converges to a stable law.

At this point, two questions (at least) arise: 1) what happens if  $f$  is no longer continuous? 2) what happens if  $f$  is no longer bounded? For instance, for the uniformly expanding map  $T_0(x) = 2x - [2x]$ , the central limit theorem holds with the normalization  $\sqrt{n}$  as soon as  $f$  is monotonic and square integrable on  $[0, 1]$ , that is not necessarily continuous nor bounded.

For the slightly different map  $\theta_\gamma(x) = x(1 - x^\gamma)^{-1/\gamma} - [x(1 - x^\gamma)^{-1/\gamma}]$ , with the same behavior around the indifferent fixed point, Raugi (2004) (following a work by Conze and Raugi (2003)) has given a precise criterion for the central limit theorem with the normalization  $\sqrt{n}$  in the case where  $0 < \gamma < 1/2$  (see his Corollary 1.7). In particular his result applies to a large class of non continuous functions, which gives a quite complete answer to our first question for the map  $\theta_\gamma$ . The result also applies to the unbounded function  $f(x) = x^{-a}$  with  $0 < a < 1/2 - \gamma$ . However, the function  $f$  is allowed to blow up near 0 only (if  $f$  tends to infinity when  $x$  tends to  $x_0 \in ]0, 1]$ , then the variation coefficient  $v(fh_\gamma, k)$ , where  $h_\gamma$  is the density of the  $\theta_\gamma$ -invariant probability, is always infinite).

We now go back to the map  $T_\gamma$ . In a short discussion after the proof of his Theorem 1.3, Gouëzel (2004) considers the case where  $f(x) = x^{-a}$ , with  $0 < a < 1 - \gamma$ . He shows that, if  $0 < a < 1/2 - \gamma$  then the central limit theorem holds with the normalization  $\sqrt{n}$ , if  $a = 1/2 - \gamma$  then the central limit theorem holds with the normalization  $\sqrt{n \ln(n)}$ , and if  $0 < a < 1 - \gamma$  and  $\gamma \geq 1/2$  then there is convergence to a stable law. Again, as for Raugi's result (2004) concerning the map  $\theta_\gamma$ , the function  $f$  is allowed to blow up only near 0.

On another hand, we know that for stationary Harris recurrent Markov chains with invariant measure  $\mu$  and  $\beta$ -mixing coefficients of order  $n^{-b}$ ,  $b > 1$ , the central limit theorem holds with the normalization  $\sqrt{n}$  as soon as the moment condition  $\mu(|f|^p) < \infty$  holds for  $p > 2b/(b - 1)$ . For  $T_\gamma$ , the covariances decay is of order  $n^{(\gamma-1)/\gamma}$ , so that one can expect the moment condition  $\nu_\gamma(|f|^p) < \infty$  for  $p > (2 - 2\gamma)/(1 - 2\gamma)$ . For instance, if  $f(x) = x^{-a}$ , since the density of  $\nu_\gamma$  is of order  $x^{-\gamma}$  near 0, the moment condition is satisfied if  $0 < a < 1/2 - \gamma$ , which is coherent with Gouëzel's result (2004). However, since the chain  $(K_\gamma, \nu_\gamma)$  is not  $\beta$ -mixing, the condition  $\nu_\gamma(|f|^p) < \infty$  for  $p > (2 - 2\gamma)/(1 - 2\gamma)$  alone is not sufficient to imply the central limit theorem, and one still needs some regularity on  $f$ .

Let us now define the class of functions of interest. For any probability measure  $\mu$  on  $\mathbb{R}$ , any  $M > 0$  and any  $p \in ]1, \infty]$ , let  $\text{Mon}(M, p, \mu)$  be the class of functions  $g$  which are monotonic on some open interval of  $\mathbb{R}$  and null elsewhere, and such that  $\mu(|g| > t) \leq M^p t^{-p}$  for  $p < \infty$  and  $\mu(|g| > M) = 0$  for  $p = \infty$ . Let  $\mathcal{C}(M, p, \mu)$  be the closure in  $\mathbb{L}^1(\mu)$  of the set of functions which can be written as  $\sum_{i=1}^n a_i g_i$ , where  $\sum_{i=1}^n |a_i| \leq 1$  and  $g_i$  belongs to  $\text{Mon}(M, p, \mu)$ . Note that a function belonging to  $\mathcal{C}(M, p, \mu)$  is allowed to blow up at an infinite number of points.

In Corollary 4.1 of the present paper, we prove that if  $f$  belongs to the class  $\mathcal{C}(M, p, \nu_\gamma)$  for  $p > (2 - 2\gamma)/(1 - 2\gamma)$ , then  $n^{-1/2}(S_n(f - \nu_\gamma(f)))$  converges in distribution to a normal law. We also give some conditions on  $p$  to obtain rates of convergence in the central limit theorem (Corollary 5.1), as well as moment inequalities for  $S_n(f - \nu_\gamma(f))$  (Corollary 6.1). Finally, a central limit theorem for the empirical distribution function of  $(T_\gamma^i)_{1 \leq i \leq n}$  is given in the last section (Corollary 7.1).

To prove these results, we compute the  $\beta$ -dependence coefficients (cf Dedecker and Prieur (2005, 2007)) of the Markov chain  $(K_\gamma, \nu_\gamma)$ . The main tool is a precise estimate of the Perron-Frobenius operator of the map  $F$  associated to  $T_\gamma$  on the Young tower, due to Maume-Deschamps (2001). Next, we apply some general results for  $\beta$ -dependent Markov chains. For the sake of simplicity, we give all

the computations in the case of the maps  $T_\gamma$ , but our arguments remain valid for many other systems modelled by Young towers.

## 2. THE MAIN INEQUALITY

For any Markov kernel  $K$  with invariant measure  $\mu$ , any non-negative integers  $n_1, n_2, \dots, n_k$ , and any bounded measurable functions  $f_1, f_2, \dots, f_k$ , define

$$\begin{aligned} K^{(n_1, n_2, \dots, n_k)}(f_1, f_2, \dots, f_k) &= K^{n_1}(f_1 K^{n_2}(f_2 K^{n_3}(f_3 \cdots K^{n_{k-1}}(f_{k-1} K^{n_k}(f_k)) \cdots))), \text{ and} \\ K^{(0)(n_1, n_2, \dots, n_k)}(f_1, f_2, \dots, f_k) &= K^{(n_1, n_2, \dots, n_k)}(f_1, f_2, \dots, f_k) - \mu(K^{(n_1, n_2, \dots, n_k)}(f_1, f_2, \dots, f_k)). \end{aligned}$$

For  $\alpha \in ]0, 1]$  and  $c > 0$ , let  $H_{\alpha, c}$  be the set of functions  $f$  such that  $|f(x) - f(y)| \leq c|x - y|^\alpha$ .

**Theorem 2.1.** *Let  $\gamma \in ]0, 1[$ , and let  $f^{(0)} = f - \nu_\gamma(f)$ . For any  $\alpha \in ]0, 1]$ , the following inequality holds:*

$$\nu_\gamma \left( \sup_{f_1, \dots, f_k \in H_{\alpha, 1}} |K_\gamma^{(0)(n_1, n_2, \dots, n_k)}(f_1^{(0)}, f_2^{(0)}, \dots, f_k^{(0)})| \right) \leq \frac{C(\alpha, k)(\ln(n_1 + 1))^2}{(n_1 + 1)^{(1-\gamma)/\gamma}}.$$

In particular,

$$\nu_\gamma \left( \sup_{f \in H_{\alpha, 1}} |K_\gamma^n f - \nu_\gamma(f)| \right) \leq \frac{C(\alpha, 1)(\ln(n + 1))^2}{(n + 1)^{(1-\gamma)/\gamma}}.$$

**Proof of Theorem 2.1.** We refer to the paper by Young (1999) for the construction of the tower  $\Delta$  associated to  $T_\gamma$  (with floors  $\Lambda_\ell$ ), and for the mappings  $\pi$  from  $\Delta$  to  $[0, 1]$  and  $F$  from  $\Delta$  to  $\Delta$  such that  $T_\gamma \circ \pi = \pi \circ F$ . On  $\Delta$  there is a probability measure  $m_0$  and an unique  $F$ -invariant probability measure  $\bar{\nu}$  with density  $h_0$  with respect to  $m_0$ , and  $\bar{\nu}(\Lambda_\ell) = O(\ell^{-1/\gamma})$ . The unique  $T_\gamma$ -invariant probability measure  $\nu_\gamma$  is then given by  $\nu_\gamma = \bar{\nu}^\pi$ . There exists a distance  $\delta$  on  $\Delta$  such that  $\delta(x, y) \leq 1$  and  $|\pi(x) - \pi(y)| \leq \kappa\delta(x, y)$ . For  $\alpha \in ]0, 1]$ , let  $\delta_\alpha = \delta^\alpha$ , let  $L_\alpha$  be the space of Lipschitz functions with respect to  $\delta_\alpha$ , and let  $L_\alpha(f) = \sup_{x, y \in \Delta} |f(x) - f(y)|/\delta_\alpha(x, y)$ . Let  $L_{\alpha, c}$  be the set of functions such that  $L_\alpha(f) \leq c$ . For  $\varphi$  in  $H_{\alpha, c}$ , the function  $\varphi \circ \pi$  belongs to  $L_{\alpha, c\kappa^\alpha}$ . Any function  $f$  in  $L_\alpha$  is bounded and the space  $L_\alpha$  is a Banach space with respect to the norm  $\|f\|_\alpha = L_\alpha(f) + \|f\|_\infty$ . The density  $h_0$  belongs to any  $L_\alpha$  and  $1/h_0$  is bounded. As in Maume-Deschamps (2001), we denote by  $\mathcal{L}_0$  the Perron-Frobenius operator of  $F$  with respect to  $m_0$ , and by  $P$  the Perron-Frobenius operator of  $F$  with respect to  $\bar{\nu}$ : for any bounded measurable functions  $\varphi, \psi$ ,

$$m_0(\varphi \cdot \psi \circ F) = m_0(\mathcal{L}_0(\varphi)\psi) \quad \text{and} \quad \bar{\nu}(\varphi \cdot \psi \circ F) = \bar{\nu}(P(\varphi)\psi).$$

We first state a useful lemma

**Lemma 2.1.** *For any positive  $n_1, n_2, \dots, n_k$  and any bounded measurable functions  $f_1, f_2, \dots, f_k$  from  $[0, 1]$  to  $\mathbb{R}$ , one has*

$$K_\gamma^{(n_1, n_2, \dots, n_k)}(f_1, f_2, \dots, f_k) \circ \pi = \mathbb{E}_{\bar{\nu}}(P^{(n_1, n_2, \dots, n_k)}(f_1 \circ \pi, f_2 \circ \pi, \dots, f_k \circ \pi) | \pi).$$

We now complete the proof of Theorem 2.1 for  $k = 2$ , the general case being similar. Applying Lemma 2.1, it follows that

$$\begin{aligned} \sup_{f, g \in H_{\alpha, 1}} |K_\gamma^n(f^{(0)} K_\gamma^m g^{(0)})(x) - \nu_\gamma(f^{(0)} K_\gamma^m g^{(0)})| \\ \leq \mathbb{E}_{\bar{\nu}} \left( \sup_{\phi, \psi \in L_{\alpha, \kappa^\alpha}} |P^n(\phi^{(0)} P^m \psi^{(0)}) - \bar{\nu}(\phi^{(0)} P^m \psi^{(0)})| | \pi = x \right). \end{aligned}$$

Here, we need the following lemma, which is derived from Lemma 3.4 in Maume-Deschamps (2001).

**Lemma 2.2.** *There exists  $M_\alpha > 0$  such that, for any  $\psi \in L_\alpha$ ,*

$$|P^m\psi(x) - P^m\psi(y)| \leq M_\alpha\delta_\alpha(x, y)\|\psi^{(0)}\|_\alpha \leq 2M_\alpha\delta_\alpha(x, y)L_\alpha(\psi).$$

Hence, if  $\psi \in L_{\alpha, \kappa^\alpha}$ , then  $P^m(\psi^{(0)})$  belongs to  $L_{\alpha, 2M_\alpha\kappa^\alpha}$  and is centered, so that  $\phi^{(0)}P^m\psi^{(0)}$  belongs to  $L_{\alpha, 4M_\alpha\kappa^{2\alpha}}$ . It follows that

$$\sup_{f, g \in H_{\alpha, 1}} |K_\gamma^n(f^{(0)}K_\gamma^m g^{(0)})(x) - \nu(f^{(0)}K_\gamma^m g^{(0)})| \leq 4M_\alpha\kappa^{2\alpha}\mathbb{E}_{\bar{\nu}}\left(\sup_{\varphi \in L_{\alpha, 1}} |P^n(\varphi) - \bar{\nu}(\varphi)| \mid \pi = x\right).$$

Next, we apply the following Lemma, which is derived from Corollary 3.14 in Maume-Deschamps (2001).

**Lemma 2.3.** *Let  $v_\ell = (\ell + 1)^{(1-\gamma)/\gamma}(\ln(\ell + 1))^{-2}$ . There exists  $C_\alpha > 0$  such that*

$$\mathbb{E}_{\bar{\nu}}\left(\sup_{\varphi \in L_{\alpha, 1}} |P^n(\varphi) - \bar{\nu}(\varphi)| \mid \pi = x\right) \leq C_\alpha(\ln(n + 1))^2(n + 1)^{(\gamma-1)/\gamma} \sum_{\ell \geq 0} v_\ell \mathbb{E}_{\bar{\nu}}(\mathbf{1}_{\Lambda_\ell} \mid \pi = x).$$

Hence

$$\nu_\gamma\left(\sup_{f, g \in H_{\alpha, 1}} |K_\gamma^n(f^{(0)}K_\gamma^m g^{(0)}) - \nu(f^{(0)}K_\gamma^m g^{(0)})|\right) \leq 4M_\alpha\kappa^{2\alpha}C_\alpha(\ln(n + 1))^2(n + 1)^{(\gamma-1)/\gamma} \sum_{\ell \geq 0} v_\ell \bar{\nu}(\Lambda_\ell).$$

Since  $\bar{\nu}(\Lambda_\ell) = O(\ell^{-1/\gamma})$ , the result follows.

**Proof of Lemma 2.1.** We write the proof for  $k = 2$  only, the general case being similar. Let  $\varphi, f$  and  $g$  be three bounded measurable functions. One has

$$\begin{aligned} \nu_\gamma(\varphi K_\gamma^n(f K_\gamma^m g)) &= \nu_\gamma(\varphi \circ T_\gamma^{n+m} \cdot f \circ T_\gamma^m \cdot g) \\ &= \bar{\nu}(\varphi \circ \pi \circ F^{n+m} \cdot f \circ \pi \circ F^m \cdot g \circ \pi) \\ &= \bar{\nu}(\varphi \circ \pi P^n(f \circ \pi P^m(g \circ \pi))) \\ &= \bar{\nu}(\varphi \circ \pi \mathbb{E}_{\bar{\nu}}(P^n(f \circ \pi P^m(g \circ \pi)) \mid \pi)) \\ &= \int \varphi(x) \mathbb{E}_{\bar{\nu}}(P^n(f \circ \pi P^m(g \circ \pi)) \mid \pi = x) \nu_\gamma(dx), \end{aligned}$$

which proves Lemma 2.1 for  $k = 2$ .

**Proof of Lemma 2.2.** Applying Lemma 3.4 in Maume-Deschamps (2001) with  $v_k = 1$ , we see that there exists  $D_\alpha > 0$  such that, for any  $\psi$  in  $L_\alpha$ ,

$$|\mathcal{L}_0^m\psi(x) - \mathcal{L}_0^m\psi(y)| \leq D_\alpha\delta_\alpha(x, y)\|\psi\|_\alpha.$$

Now  $P^m(\psi) = \mathcal{L}_0^m(\psi h_0)/h_0$ . Since  $1/h_0$  is bounded by  $B(h_0)$ , and since  $h_0$  belongs to  $L_\alpha$ , it follows that

$$|P^m\psi(x) - P^m\psi(y)| \leq D_\alpha B(h_0)\|h_0\|_\alpha \delta_\alpha(x, y)\|\psi\|_\alpha.$$

Let  $M_\alpha = D_\alpha B(h_0)\|h_0\|_\alpha$ . Since  $|P^m\psi(x) - P^m\psi(y)| = |P^m\psi^{(0)}(x) - P^m\psi^{(0)}(y)|$  and since  $\|\psi^{(0)}\|_\infty \leq L_\alpha(\psi)$ , it follows that

$$|P^m\psi(x) - P^m\psi(y)| \leq M_\alpha\delta_\alpha(x, y)\|\psi^{(0)}\|_\alpha \leq 2M_\alpha\delta_\alpha(x, y)L_\alpha(\psi).$$

**Proof of Lemma 2.3.** Applying Corollary 3.14 in Maume-Deschamps (2001), there exists  $B_\alpha > 0$  such that

$$|\mathcal{L}_0^n f - h_0 m_0(f)| \leq B_\alpha \|f\|_\alpha (\ln(n+1))^2 (n+1)^{(\gamma-1)/\gamma} \sum_{\ell \geq 0} v_\ell \mathbf{1}_{\Delta_\ell}.$$

It follows that, with the notations of the proof of Lemma 2.2,

$$|P^n(f) - \bar{\nu}(f)| \leq B_\alpha B(h_0) \|h_0\|_\alpha \|f\|_\alpha (\ln(n+1))^2 (n+1)^{(\gamma-1)/\gamma} \sum_{\ell \geq 0} v_\ell \mathbf{1}_{\Delta_\ell}.$$

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

$$|P^n(f) - \bar{\nu}(f)| \leq 2B_\alpha B(h_0) \|h_0\|_\alpha L_\alpha(f) (\ln(n+1))^2 (n+1)^{(\gamma-1)/\gamma} \sum_{\ell \geq 0} v_\ell \mathbf{1}_{\Delta_\ell},$$

and the result follows.

### 3. THE DEPENDENCE COEFFICIENTS

Let  $\mathbf{X} = (X_i)_{i \geq 0}$  be a stationary Markov chain with invariant measure  $\mu$  and transition kernel  $K$ . Let  $f_t(x) = \mathbf{1}_{x \leq t}$ . As in Dedecker and Priour (2005, 2007), define the coefficients  $\alpha_k(n)$  of the stationary Markov chain  $(X_i)_{i \geq 0}$  by

$$\begin{aligned} \alpha_1(n) &= \sup_{t \in \mathbb{R}} \mu(|K^n(f_t) - \mu(f_t)|), \quad \text{and for } k \geq 2, \\ \alpha_k(n) &= \alpha_1(n) \vee \sup_{2 \leq l \leq k} \sup_{n_2 \geq 1, \dots, n_l \geq 1} \sup_{t_1, \dots, t_l \in \mathbb{R}} \mu(|K^{(0)(n, n_2, \dots, n_l)}(f_{t_1}, f_{t_2}, \dots, f_{t_l})|). \end{aligned}$$

In the same way, define the coefficients  $\beta_k(n)$  by

$$\begin{aligned} \beta_1(n) &= \mu\left(\sup_{t \in \mathbb{R}} |K^n(f_t) - \mu(f_t)|\right), \quad \text{and for } k \geq 2, \\ \beta_k(n) &= \beta_1(n) \vee \sup_{2 \leq l \leq k} \sup_{n_2 \geq 1, \dots, n_l \geq 1} \mu\left(\sup_{t_1, \dots, t_l \in \mathbb{R}} |K^{(0)(n, n_2, \dots, n_l)}(f_{t_1}, f_{t_2}, \dots, f_{t_l})|\right). \end{aligned}$$

**Theorem 3.1.** *Let  $0 < \gamma < 1$ . Let  $\mathbf{X} = (X_i)_{i \geq 0}$  be a stationary Markov chain with invariant measure  $\nu_\gamma$  and transition kernel  $K_\gamma$ . There exist two positive constants  $C_1(\gamma)$  and  $C_2(\delta, \gamma, k)$  such that, for any  $\delta$  in  $]0, (1-\gamma)/\gamma[$  and any positive integer  $k$ ,*

$$C_1(\gamma)(n+1)^{\frac{\gamma-1}{\gamma}} \leq \alpha_k(n) \leq \beta_k(n) \leq C_2(\delta, \gamma, k)(n+1)^{\frac{\gamma-1}{\gamma} + \delta}.$$

**Proof of Theorem 3.1.** Applying Proposition 2, Item 2, in Dedecker and Priour (2005), we know that

$$\nu_\gamma\left(\sup_{f \in H_{1,1}} |K_\gamma^n f - \nu_\gamma(f)|\right) \leq 2\alpha_1(n).$$

Hence, for any  $\varphi$  such that  $|\varphi| \leq 1$  and any  $f$  in  $H_{1,1}$ ,

$$\nu_\gamma(\varphi \cdot (K_\gamma^n f - \nu_\gamma(f))) = \nu_\gamma(\varphi \circ T^n \cdot (f - \nu_\gamma(f))) \leq 2\alpha_1(n)$$

The lower bound for  $\alpha_k(n)$  follows from the lower bound for  $\nu_\gamma(\varphi \circ T^n \cdot (f - \nu_\gamma(f)))$  given by Sarig (2002), Corollary 1.

It remains to prove the upper bound. The point is to approximate the indicator  $f_t(x) = \mathbf{1}_{x \leq t}$  by some  $\alpha$ -Hölder function. Let

$$f_{t,\epsilon,\alpha}(x) = f_t(x) + \left(1 - \left(\frac{x-t}{\epsilon}\right)^\alpha\right) \mathbf{1}_{t < x \leq t+\epsilon}.$$

This function is  $\alpha$ -Hölder with Hölder constant  $\epsilon^{-\alpha}$ . We now prove the upper bounds for  $k = 1$  and  $k = 2$  only, the general case being similar. For  $k = 1$ , one has

$$K_\gamma^n(f_{t-\epsilon,\epsilon,\alpha}) - \nu_\gamma(f_{t-\epsilon,\epsilon,\alpha}) - \nu_\gamma([t-\epsilon, t]) \leq K_\gamma^n(f_t) - \nu_\gamma(f_t) \leq K_\gamma^n(f_{t,\epsilon,\alpha}) - \nu_\gamma(f_{t,\epsilon,\alpha}) + \nu_\gamma([t, t+\epsilon]).$$

Since the density  $g_{\nu_\gamma}$  of  $\nu_\gamma$  is such that  $g_{\nu_\gamma}(x) \leq V(\gamma)x^{-\gamma}$ , we infer that for any real  $a$ ,  $\nu_\gamma([a, a+\epsilon]) \leq V(\gamma)\epsilon^{1-\gamma}(1-\gamma)^{-1}$ . Consequently,

$$|K_\gamma^n(f_t) - \nu_\gamma(f_t)| \leq \epsilon^{-\alpha} \sup_{f \in H_{\alpha,1}} |K_\gamma^n(f) - \nu_\gamma(f)| + \frac{V(\gamma)}{1-\gamma} \epsilon^{1-\gamma}.$$

Applying Theorem 2.1 with  $k = 1$ , we obtain that

$$\nu_\gamma \left( \sup_{t \in [0,1]} |K_\gamma^n(f_t) - \nu_\gamma(f_t)| \right) \leq C(\alpha, 1) \epsilon^{-\alpha} (\ln(n+1))^2 (n+1)^{\frac{\gamma-1}{\gamma}} + \frac{V(\gamma)}{1-\gamma} \epsilon^{1-\gamma}.$$

The optimal  $\epsilon$  is equal to

$$\epsilon = \left( \frac{\alpha C(\alpha, 1) (\ln(n+1))^2 (n+1)^{\frac{\gamma-1}{\gamma}}}{V(\gamma)} \right)^{\frac{1}{\alpha+1-\gamma}}.$$

Consequently, for some positive constant  $D(\gamma, \alpha)$ , one has

$$\nu_\gamma \left( \sup_{t \in [0,1]} |K_\gamma^n(f_t) - \nu_\gamma(f_t)| \right) \leq D(\gamma, \alpha) \left( (\ln(n+1))^2 (n+1)^{\frac{\gamma-1}{\gamma}} \right)^{\frac{1-\gamma}{\alpha+1-\gamma}}.$$

Choosing  $\alpha < \delta\gamma(1-\gamma)/(1-\gamma(1+\delta))$ , the result follows for  $k = 1$ .

We now prove the result for  $k = 2$ . Clearly, the four following inequalities hold:

$$\begin{aligned} K_\gamma^n(f_t^{(0)} K_\gamma^m f_s^{(0)}) &\leq K_\gamma^n(f_{t,\epsilon,\alpha}^{(0)} K_\gamma^m f_{s,\epsilon,\alpha}^{(0)}) + \nu_\gamma([t, t+\epsilon]) + \nu_\gamma([s, s+\epsilon]), \\ K_\gamma^n(f_t^{(0)} K_\gamma^m f_s^{(0)}) &\geq K_\gamma^n(f_{t-\epsilon,\epsilon,\alpha}^{(0)} K_\gamma^m f_{s-\epsilon,\epsilon,\alpha}^{(0)}) - \nu_\gamma([t-\epsilon, t]) - \nu_\gamma([s-\epsilon, s]), \\ \nu_\gamma(f_t^{(0)} K_\gamma^m f_s^{(0)}) &\geq \nu_\gamma(f_{t,\epsilon,\alpha}^{(0)} K_\gamma^m f_{s,\epsilon,\alpha}^{(0)}) - 2\nu_\gamma([t, t+\epsilon]) - \nu_\gamma([s, s+\epsilon]), \\ \nu_\gamma(f_t^{(0)} K_\gamma^m f_s^{(0)}) &\leq \nu_\gamma(f_{t-\epsilon,\epsilon,\alpha}^{(0)} K_\gamma^m f_{s-\epsilon,\epsilon,\alpha}^{(0)}) + 2\nu_\gamma([t-\epsilon, t]) + \nu_\gamma([s-\epsilon, s]). \end{aligned}$$

Consequently,

$$|K_\gamma^n(f_t^{(0)} K_\gamma^m f_s^{(0)}) - \nu_\gamma(f_t^{(0)} K_\gamma^m f_s^{(0)})| \leq \epsilon^{-\alpha} \sup_{f,g \in H_{\alpha,1}} |K_\gamma^n(f^{(0)} K_\gamma^m g^{(0)}) - \nu_\gamma(f^{(0)} K_\gamma^m g^{(0)})| + \frac{5V(\gamma)}{1-\gamma} \epsilon^{1-\gamma}.$$

Applying Theorem 2.1, we obtain that

$$\nu_\gamma \left( \sup_{t \in [0,1]} |K_\gamma^n(f_t^{(0)} K_\gamma^m f_s^{(0)}) - \nu_\gamma(f_t^{(0)} K_\gamma^m f_s^{(0)})| \right) \leq C(\alpha, 2) \epsilon^{-\alpha} (\ln(n+1))^2 (n+1)^{\frac{\gamma-1}{\gamma}} + \frac{5V(\gamma)}{1-\gamma} \epsilon^{1-\gamma},$$

and the proof can be completed as for  $k = 1$ .

## 4. CENTRAL LIMIT THEOREMS

In this section we give a central limit theorem for  $S_n(f - \nu_\gamma(f))$  when  $f$  belongs to the class  $\mathcal{C}(M, p, \mu)$  defined in the introduction. Note that any function  $f$  with bounded variation (BV) such that  $|f| \leq M_1$  and  $\|df\| \leq M_2$  belongs to the class  $\mathcal{C}(M_1 + 2M_2, \infty, \mu)$ . Hence, any BV function  $f$  belongs to  $\mathcal{C}(M, \infty, \mu)$  for some  $M$  large enough. If  $g$  is monotonic on some open interval of  $\mathbb{R}$  and null elsewhere, and if  $\mu(|g|^p) \leq M^p$ , then  $g$  belongs to  $\text{Mon}(M, p, \mu)$ . Conversely, any function in  $\mathcal{C}(M, p, \mu)$  belongs to  $\mathbb{L}^q(\mu)$  for  $1 \leq q < p$ .

**Theorem 4.1.** *Let  $\mathbf{X} = (X_i)_{i \geq 0}$  be a stationary and ergodic (in the ergodic theoretic sense) Markov chain with invariant measure  $\mu$  and transition kernel  $K$ . Assume that  $f$  belongs to  $\mathcal{C}(M, p, \mu)$  for some  $M > 0$  and some  $p \in ]2, \infty]$ , and that*

$$\sum_{k>0} (\alpha_1(k)) \frac{p-2}{p} < \infty.$$

The following results hold:

- (1) *The series*

$$\sigma^2(\mu, K, f) = \mu((f - \mu(f))^2) + 2 \sum_{k>0} \mu((f - \mu(f))K^k(f))$$

*converges to some non negative constant, and  $n^{-1} \text{Var}(\sum_{i=1}^n f(X_i))$  converges to  $\sigma^2(\mu, K, f)$ .*

- (2) *Let  $(D([0, 1], d))$  be the space of cadlag functions from  $[0, 1]$  to  $\mathbb{R}$  equipped with the Skorohod metric  $d$ . The process  $\{n^{-1/2} \sum_{i=1}^{[nt]} (f(X_i) - \mu(f)), t \in [0, 1]\}$  converges in distribution in  $(D([0, 1], d))$  to  $\sigma(\mu, K, f)W$ , where  $W$  is a standard Wiener process.*
- (3) *One has the representation*

$$f(X_1) - \mu(f) = m(X_1, X_0) + g(X_1) - g(X_0)$$

*with  $\mu(|g|^{p/(p-1)}) < \infty$ ,  $\mathbb{E}(m(X_1, X_0)|X_0) = 0$  and  $\mathbb{E}(m^2(X_1, X_0)) = \sigma^2(\mu, K, f)$ .*

**Corollary 4.1.** *Let  $\gamma \in ]0, 1/2[$ . If  $f$  belongs to the class  $\mathcal{C}(M, p, \nu)$  for some  $M > 0$  and some  $p > (2 - 2\gamma)/(1 - 2\gamma)$ , then  $n^{-1/2} S_n(f - \nu_\gamma(f))$  converges in distribution to  $\mathcal{N}(0, \sigma^2(\nu_\gamma, K_\gamma, f))$ .*

**Remark 4.1.** *We infer from Corollary (4.1) that the central limit theorem holds for any BV function provided  $\gamma < 1/2$ . Under the same condition on  $\gamma$ , Young (1999) has proved that the central limit theorem holds for any  $\alpha$ -Hölder function. For the map  $\theta_\gamma(x) = x(1 - x^\gamma)^{-1/\gamma} - [x(1 - x^\gamma)^{-1/\gamma}]$  and  $\gamma < 1/2$ , the central limit theorem for BV functions is a consequence of Corollary 1.7(i) in Raugi (2004).*

**Two simple examples.**

- (1) Assume that  $f$  is positive and non increasing on  $]0, 1[$ , with  $f(x) \leq Cx^{-a}$  for some  $a \geq 0$ . Since the density  $g_{\nu_\gamma}$  of  $\nu_\gamma$  is such that  $g_{\nu_\gamma}(x) \leq V(\gamma)x^{-\gamma}$ , we infer that

$$\nu_\gamma(f > t) \leq \frac{C \frac{1-\gamma}{a} V(\gamma)}{1-\gamma} t^{-\frac{1-\gamma}{a}}.$$

Hence the CLT holds as soon as  $a < \frac{1}{2} - \gamma$ .

(2) Assume now that  $f$  is positive and non decreasing on  $]0, 1[$  with  $f(x) \leq C(1-x)^{-a}$  for some  $a \geq 0$ . Here

$$\nu_\gamma(f > t) \leq \frac{V(\gamma)}{1-\gamma} \left(1 - \left(1 - \left(\frac{C}{t}\right)^{1/a}\right)^{1-\gamma}\right).$$

Hence the CLT holds as soon as  $a < \frac{1}{2} - \frac{\gamma}{2(1-\gamma)}$ .

**Proof of Theorem 4.1.** Let  $f$  in  $\mathcal{C}(M, p, \mu)$ . From Dedecker and Rio (2000), Items (1) and (2) of Theorem 4.1 hold as soon as

$$\sum_{n>0} \|(f(X_0) - \mu(f))(\mathbb{E}(f(X_n)|X_0) - \mu(f))\|_1 < \infty.$$

Assume first that  $f = \sum_{i=1}^k a_i g_i$ , where  $\sum_{i=1}^k |a_i| \leq 1$ , and  $g_i$  belongs to  $\text{Mon}(M, p, \mu)$ . Clearly, the series on left side is bounded by

$$\sum_{i=1}^k \sum_{j=1}^k |a_i a_j| \sum_{n>0} \|(g_i(X_0) - \mu(g_i))(\mathbb{E}(g_j(X_n)|X_0) - \mu(g_j))\|_1.$$

Here, we use the following lemma

**Lemma 4.1.** *Let  $g_i$  and  $g_j$  be two functions in  $\text{Mon}(M, p, \mu)$  for some  $p \in ]2, \infty]$ . For any  $1 \leq q \leq p$  one has*

$$\|\mathbb{E}(g_j(X_n)|X_0) - \mu(g_j)\|_q \leq 2M \left(\frac{p}{p-q}\right)^{1/q} (2\alpha_1(n))^{\frac{p-q}{pq}}.$$

For any  $1 \leq q < p/2$ , one has

$$\|(g_i(X_0) - \mu(g_i))(\mathbb{E}(g_j(X_n)|X_0) - \mu(g_j))\|_q \leq 4M^2 \left(\frac{p}{p-2q}\right)^{1/q} (2\alpha_1(n))^{\frac{p-2q}{pq}}.$$

From Lemma 4.1 with  $q = 1$ , we conclude that

$$(4.1) \quad \sum_{n>0} \|(f(X_0) - \mu(f))(\mathbb{E}(f(X_n)|X_0) - \mu(f))\|_1 \leq \frac{4pM^2}{p-2} \sum_{n>0} (2\alpha_1(n))^{\frac{p-2}{p}}.$$

Since the bound (4.1) is true for any function  $f = \sum_{i=1}^k a_i g_i$ , it is true also for any  $f$  in  $\mathcal{C}(M, p, \mu)$ , and Items (1) and (2) follow.

The last assertion is rather standard. From the first inequality of Lemma 4.1 with  $q = p/(p-1)$ , we infer that if  $\sum_{n>0} (\alpha_1(n))^{(p-2)/p} < \infty$ , then  $\sum_{n>0} \|\mathbb{E}(f(X_n)|X_0) - \mu(f)\|_{p/(p-1)} < \infty$  for any  $f$  in  $\mathcal{C}(M, p, \mu)$ . It follows that  $g(x) = \sum_{k=1}^{\infty} \mathbb{E}(f(X_k) - \mu(f)|X_0 = x)$  belongs to  $\mathbb{L}^{p/(p-1)}(\mu)$  and that  $m(X_1, X_0) = \sum_{k \geq 1} (\mathbb{E}(f(X_k)|X_0) - \mathbb{E}(f(X_k)|X_1))$  belongs to  $\mathbb{L}^{p/(p-1)}$ . Clearly

$$f(X_1) - \mu(f) = m(X_1, X_0) + g(X_0) - g(X_1),$$

with  $\mathbb{E}(m(X_1, X_0)|X_0) = 0$ . Moreover, it follows from the preceding result that

$$\lim_{n \rightarrow \infty} \frac{1}{\sqrt{n}} \left\| \sum_{k=1}^n m(X_k, X_{k-1}) \right\|_1 = \lim_{n \rightarrow \infty} \frac{1}{\sqrt{n}} \left\| \sum_{k=1}^n (f(X_k) - \mu(f)) \right\|_1 \leq \sigma(\mu, K, f).$$

By Theorem 1 in Esseen and Janson (1985), it follows that  $\mathbb{E}(m^2(X_1, X_0)) = \sigma^2(\mu, K, f)$ .



**Proof of Lemma 4.1.** We only prove the second inequality (the proof of the first one is easier). Let  $r = q/(q-1)$  and let  $B_r(\sigma(X_0))$  be the set of  $\sigma(X_0)$ -measurable random variables such that  $\|Y\|_r \leq 1$ . By duality,

$$\begin{aligned} \|(g_i(X_0) - \mu(g_i))(\mathbb{E}(g_j(X_n)|X_0) - \mu(g_j))\|_q &= \sup_{Y \in B_r(\sigma(X_0))} \mathbb{E}(Y(g_i(X_0) - \mu(g_i))(g_j(X_n) - \mu(g_j))) \\ &= \sup_{Y \in B_r(\sigma(X_0))} \text{Cov}(Y(g_i(X_0) - \mu(g_i)), g_j(X_n)). \end{aligned}$$

Define the coefficients  $\alpha_{k,g}(n)$  of the sequence  $(g(X_i))_{i \geq 0}$  as in Section 3 with  $g \circ f_t$  instead of  $f_t$ . If  $g$  is monotonic on some open interval of  $\mathbb{R}$  and null elsewhere, the set  $\{x : g(x) \leq t\}$  is either some interval or the complement of some interval, so that  $\alpha_{k,g}(n) \leq 2^k \alpha_k(n)$ . Let  $Q_Y$  be the generalized inverse of the tail function  $t \rightarrow \mathbb{P}(|Y| > t)$ . From Theorem 1.1 and Lemma 2.1 in Rio (2000), one has that

$$\begin{aligned} \text{Cov}(Y g_i(X_0), g_j(X_n)) &\leq 2 \int_0^{\alpha_{1,g_i}(n)} Q_Y(u) Q_{g_i(X_0)}(u) Q_{g_j(X_0)}(u) du \\ &\leq 2 \int_0^{2\alpha_1(n)} Q_Y(u) Q_{g_i(X_0)}(u) Q_{g_j(X_0)}(u) du. \end{aligned}$$

In the same way, applying first Theorem 1.1 in Rio (2000) and next Fréchet's inequality (1957) (see also Inequality (1.11b) in Rio (2000)),

$$\begin{aligned} \text{Cov}(Y \mu(g_i), g_j(X_n)) &\leq 2\mu(|g_i|) \int_0^{2\alpha_1(n)} Q_Y(u) Q_{g_j(X_0)}(u) du \\ &\leq 2 \int_0^{2\alpha_1(n)} Q_Y(u) Q_{g_i(X_0)}(u) Q_{g_j(X_0)}(u) du. \end{aligned}$$

Since  $\int_0^1 Q_Y^r(u) du \leq 1$ , it follows that

$$\|(g_i(X_0) - \mu(g_i))(\mathbb{E}(g_j(X_n)|X_0) - \mu(g_j))\|_q \leq 4 \left( \int_0^{2\alpha_1(n)} Q_{g_i(X_0)}^q(u) Q_{g_j(X_0)}^q(u) du \right)^{1/q}.$$

Since  $g_i$  and  $g_j$  belong to  $\text{Mon}(M, p, \mu)$  for some  $p > 2q$ , we have that  $Q_{g_i(X_0)}(u)$  and  $Q_{g_j(X_0)}(u)$  are smaller than  $Mu^{-1/p}$ , and the result follows.

**Proof of Corollary 4.1.** We have seen that  $(T_\gamma^1, \dots, T_\gamma^n)$  is distributed as  $(X_n, \dots, X_1)$  where  $(X_i)_{i \geq 0}$  is the stationary Markov chain with invariant measure  $\nu_\gamma$  and transition kernel  $K_\gamma$ . Consequently, on the probability space  $([0, 1], \nu_\gamma)$ , the sum  $S_n(f - \nu_\gamma(f))$  is distributed as  $\sum_{i=1}^n (f(X_i) - \nu_\gamma(f))$ , so that  $n^{-1/2} S_n(f - \nu_\gamma(f))$  satisfies the central limit theorem if and only if  $n^{-1/2} \sum_{i=1}^n (f(X_i) - \nu_\gamma(f))$  does. Moreover, we infer from Theorem 3.1 that

$$\alpha_1(n) = O(n^{\frac{\gamma-1}{\gamma} + \epsilon})$$

for any  $\epsilon > 0$ . Consequently, if  $p > (2 - 2\gamma)/(1 - 2\gamma)$ , one has that  $\sum_{k>0} (\alpha_1(n))^{\frac{p-2}{p}} < \infty$  so that Theorem 4.1 applies: the central limit theorem holds provided that  $f$  belongs to  $\mathcal{C}(M, p, \nu_\gamma)$ .

## 5. RATES OF CONVERGENCE IN THE CLT

Let  $c$  be some concave function from  $\mathbb{R}^+$  to  $\mathbb{R}^+$ , with  $c(0) = 0$ . Denote by  $\text{Lip}_c$  the set of functions  $g$  such that

$$|g(x) - g(y)| \leq c(|x - y|).$$

When  $c(x) = x^\alpha$  for  $\alpha \in ]0, 1]$ , we have  $\text{Lip}_c = H_{\alpha,1}$ . For two probability measures  $P, Q$  with finite first moment, let

$$d_c(P, Q) = \sup_{g \in \text{Lip}_c} |P(g) - Q(g)|.$$

When  $c = \text{Id}$ , we write  $d_c = d_1$ . Note that  $d_1(P, Q)$  is the so-called Kantorovič distance between  $P$  and  $Q$ .

**Theorem 5.1.** *Let  $\mathbf{X} = (X_i)_{i \geq 0}$  be a stationary Markov chain with invariant measure  $\mu$  and transition kernel  $K$ . Let  $\sigma^2(f) = \sigma^2(\mu, K, f)$  be the non-negative number defined in Theorem 4.1, and let  $G_{\sigma^2(f)}$  be the Gaussian distribution with mean 0 and variance  $\sigma^2(f)$ . Let  $P_n(f)$  be the distribution of the normalized sum  $n^{-1/2} \sum_{i=1}^n (f(X_i) - \mu(f))$ .*

- (1) *Assume that  $f$  belongs to  $\mathcal{C}(M, p, \mu)$  for some  $M > 0$  and some  $p \in ]2, \infty]$ , and that*

$$\sum_{k>0} (\alpha_1(k))^{p-2} < \infty.$$

*If  $\sigma^2(f) = 0$ , then  $d_c(P_n(f), \delta_{\{0\}}) = O(c(n^{-1/2}))$ .*

- (2) *If  $f$  belongs to  $\mathcal{C}(M, p, \mu)$  for some  $M > 0$  and some  $p \in ]3, \infty]$ , and if*

$$\sum_{k>0} k(\alpha_3(k))^{p-3} < \infty,$$

*then  $d_c(P_n(f), G_{\sigma^2(f)}) = O(c(n^{-1/2}))$ .*

- (3) *If  $f$  belongs to  $\mathcal{C}(M, p, \mu)$  for some  $M > 0$  and some  $p \in ]3, \infty]$ , and if*

$$\alpha_2(k) = O(k^{-(1+\delta)p/(p-3)}) \quad \text{for some } \delta \in ]0, 1[,$$

*then  $d_c(P_n(f), G_{\sigma^2(f)}) = O(c(n^{-\delta/2}))$ .*

**Corollary 5.1.** *Let  $\delta \in ]0, 1]$  and  $\gamma < 1/(2+\delta)$ , and let  $\mu_n(f)$  be the distribution of  $n^{-1/2} S_n(f - \nu_\gamma(f))$ . If  $f$  belongs to the class  $\mathcal{C}(M, p, \nu_\gamma)$  for some  $M > 0$  and some  $p > (3 - 3\gamma)/(1 - (2 + \delta)\gamma)$ , then  $d_c(\mu_n(f), G_{\sigma^2(f)}) = O(c(n^{-\delta/2}))$ , where  $\sigma^2(f) = \sigma^2(\nu_\gamma, K_\gamma, f)$ .*

**Remark 5.1.** *We infer from Corollary 5.1 that if  $f$  is BV, then  $d_1(\mu_n(f), G_{\sigma^2(f)}) = O(n^{-1/2})$  if  $\gamma < 1/3$ , and  $d_1(\mu_n(f), G_{\sigma^2(f)}) = O(n^{-\delta/2})$  if  $\gamma < 1/(2 + \delta)$ . Denote by  $d_{BV}(P, Q)$  the uniform distance between the distribution functions of  $P$  and  $Q$ . If  $f$  is  $\alpha$ -Hölder, Gouëzel (2005, Theorem 1.5) has proved that  $d_{BV}(\mu_n(f), G_{\sigma^2(f)}) = O(n^{-1/2})$  if  $\gamma < 1/3$ , and  $d_{BV}(\mu_n(f), G_{\sigma^2(f)}) = O(n^{-\delta/2})$  if  $\gamma = 1/(2 + \delta)$ . In fact, from a general result of Bolthausen (1982) for Harris recurrent Markov chains, we conjecture that the results of Corollary 5.1 are true with  $d_{BV}$  instead of  $d_1$ .*

**Two simple examples (continued).**

- (1) *Assume that  $f$  is positive and non increasing on  $[0, 1]$ , with  $f(x) \leq Cx^{-a}$  for some  $a \geq 0$ . Let  $\delta \in ]0, 1]$  and  $\gamma < 1/(2 + \delta)$ . If  $a < \frac{1}{3} - \frac{(2+\delta)\gamma}{3}$ , then  $d_c(\mu_n(f), G_{\sigma^2(f)}) = O(c(n^{-\delta/2}))$ .*

- (2) Assume that  $f$  is positive and non increasing on  $[0, 1]$ , with  $f(x) \leq C(1-x)^{-a}$  for some  $a \geq 0$ . Let  $\delta \in ]0, 1]$  and  $\gamma < 1/(2 + \delta)$ . If  $a < \frac{1}{3} - \frac{(1+\delta)\gamma}{3(1-\gamma)}$ , then  $d_c(\mu_n(f), G_{\sigma^2}(f)) = O(c(n^{-\delta/2}))$ .

**Proof of Theorem 5.1.** From the Kantorovič-Rubinstěin theorem (1957), there exists a probability measure  $\pi$  with margins  $P$  and  $Q$ , such that  $d_1(P, Q) = \int |x - y|\pi(dx, dy)$ . Since  $c$  is concave, we then have

$$d_c(P, Q) = \sup_{f \in H_c} \left| \int (f(x) - f(y))\pi(dx, dy) \right| \leq \int c(|x - y|)\pi(dx, dy) \leq c(d_1(P, Q)).$$

Hence, it is enough to prove the theorem for  $d_1$  only.

If  $\sum_{k>0} (\alpha_1(k))^{(p-2)/p} < \infty$ ,  $f$  belongs to  $\mathcal{C}(M, p, \mu)$  for some  $M > 0$  and some  $p \in ]2, \infty]$ , and  $\sigma^2(f) = 0$ , it follows from Theorem 4.1 that  $f(X_1) = g(X_0) - g(X_1)$  with  $\mu(|g|) < \infty$ . Hence

$$d_1(P_n(f), \delta_{\{0\}}) \leq \frac{2\mu(|g|)}{\sqrt{n}},$$

and Item (1) is proved.

From now, we assume that  $\sigma^2(f) > 0$  (otherwise, the result follows from Item (1)). If  $f = g_1 - g_2$ , where  $g_1, g_2$  belong to  $\text{Mon}(M, p, \mu)$  for some  $M > 0$  and some  $p \in ]3, \infty]$ , Item (2) of Theorem 5.1 follows from Theorem 3.1(b) in Dedecker and Rio (2007). In fact the proof remains unchanged if  $f$  belongs to  $\mathcal{C}(M, p, \mu)$  for some  $M > 0$  and some  $p \in ]3, \infty]$ .

It remains to prove Item (3). Let  $Y_k = f(X_k) - \mu(f)$ ,  $\sigma^2(f) = \sigma^2$ , and  $s_m = \sum_{i=1}^m Y_i$ . Define

$$W_m = A_m + B_m, \quad \text{with} \quad A_m = \mathbb{E}(s_m^2 | X_0) - m\sigma^2 \quad \text{and} \quad B_m = 2 \sum_{k=1}^m \mathbb{E}\left(Y_k \sum_{i>m} Y_i \middle| X_0\right).$$

From Theorem 2.2 in Dedecker and Rio (2007), we have that, if  $\sum_{k>0} \|Y_0 \mathbb{E}(Y_k | X_0)\|_1 < \infty$ ,

$$(5.2) \quad \sqrt{n}d_1(P_n(f), G_{\sigma^2}) \leq C \ln(n) + \sum_{m=1}^{\lfloor \sqrt{2n} \rfloor} \frac{\|(|Y_0| + 2\sigma)W_m\|_1}{m\sigma^2} + D_{1,n} + D_{2,n},$$

where

$$D_{1,n} = \sum_{m=1}^n \frac{1}{\sigma\sqrt{m}} \sum_{i \geq m} \|Y_0 \mathbb{E}(Y_i | X_0)\|_1 \quad \text{and} \quad D_{2,n} = \sum_{m=1}^n \frac{1}{2\sigma^2 m} \sum_{k=1}^m \|(\sigma^2 + Y_0^2) \mathbb{E}(Y_k | X_0)\|_1.$$

From Lemma 4.1 with  $q = 1$ , the bound (4.1) holds for any  $f$  in  $\mathcal{C}(M, p, \mu)$  for  $p > 2$ . Consequently, if  $\alpha_2(k) = O(k^{-(1+\delta)p/(p-3)})$  for some  $\delta \in ]0, 1[$  and  $p > 3$ , then  $\sum_{k>0} \|Y_0 \mathbb{E}(Y_k | X_0)\|_1 < \infty$ , so that the bound (5.2) holds. Moreover  $n^{-1/2}D_{1,n} = O(n^{-1/2} \ln(n) \vee n^{-\delta})$ . Arguing as in Lemma 4.1, one can prove that

$$\|Y_0^2 \mathbb{E}(Y_k | X_0)\|_1 \leq C(M, p)(\alpha_1(k))^{\frac{p-3}{p}},$$

so that  $n^{-1/2}D_{2,n} = O(n^{-1/2} \ln(n))$ .

Arguing as in Lemma 4.1, one can prove that, for  $0 < k < i$ ,

$$(5.3) \quad \|(|Y_0| + 2\sigma)\mathbb{E}(Y_k Y_i | X_0)\|_1 \leq \|(|Y_0| + 2\sigma)Y_k \mathbb{E}(Y_i | X_k)\|_1 \leq C(M, p, \sigma)(\alpha_1(i-k))^{\frac{p-3}{p}}.$$

Consequently,

$$\frac{1}{\sqrt{n}} \sum_{m=1}^{\lfloor \sqrt{2n} \rfloor} \frac{\|(|Y_0| + 2\sigma)B_m\|_1}{m\sigma^2} = O\left(\frac{1}{\sqrt{n}} \sum_{m=1}^{\lfloor \sqrt{2n} \rfloor} \frac{1}{m\sigma^2} \sum_{k=1}^m \sum_{i>m} \frac{1}{(i-k)^{1+\delta}}\right) = O(n^{-\delta/2}).$$

Now,

$$\frac{\|(|Y_0| + 2\sigma)A_m\|_1}{m} \leq \frac{2}{m} \sum_{i=1}^m \sum_{j=i}^m \|(|Y_0| + 2\sigma)(\mathbb{E}(Y_i Y_j | X_0) - \mathbb{E}(Y_i Y_j))\|_1 + (\|Y_0\|_1 + 2\sigma) \left| \frac{1}{m} \mathbb{E}(s_m^2) - \sigma^2 \right|.$$

For the second term on right hand, we have

$$\left| \frac{1}{m} \mathbb{E}(s_m^2) - \sigma^2 \right| \leq 2 \sum_{k=1}^{\infty} \frac{k \wedge m}{m} |\mathbb{E}(Y_0 Y_k)| = O\left(\sum_{k>0} \frac{k \wedge m}{m} (\alpha_1(k))^{\frac{p-2}{p}}\right) = O(m^{-\delta}),$$

so that

$$\frac{1}{\sqrt{n}} \sum_{m=1}^{\lfloor \sqrt{2n} \rfloor} \left| \frac{1}{m} \mathbb{E}(s_m^2) - \sigma^2 \right| = O(n^{-\delta/2}).$$

To complete the proof of the theorem, it remains to prove that

$$(5.4) \quad \frac{1}{\sqrt{n}} \sum_{m=1}^{\lfloor \sqrt{2n} \rfloor} \frac{2}{m} \sum_{i=1}^m \sum_{j=i}^m \|(|Y_0| + 2\sigma)(\mathbb{E}(Y_i Y_j | X_0) - \mathbb{E}(Y_i Y_j))\|_1 = O(n^{-\delta/2}).$$

Applying first (5.3), we have for  $j > i$ ,

$$(5.5) \quad \|(|Y_0| + 2\sigma)(\mathbb{E}(Y_i Y_j | X_0) - \mathbb{E}(Y_i Y_j))\|_1 \leq 2C(M, p, \sigma) (\alpha_1(j-i))^{\frac{p-3}{p}}.$$

We need a second bound for this quantity. Assume first that  $f = \sum_{i=1}^k a_i g_i$ , where  $\sum_{i=1}^k |a_i| \leq 1$  and  $g_i$  belongs to  $\text{Mon}(M, p, \mu)$ . Let  $g_i^{(0)} = g_i - \mu(g_i)$ . We have that

$$\begin{aligned} & \|Y_0(\mathbb{E}(Y_i Y_j | X_0) - \mathbb{E}(Y_i Y_j))\|_1 \\ & \leq \sum_{l=1}^k \sum_{q=1}^k \sum_{r=1}^k |a_l a_q a_r| \|g_l^{(0)}(X_0)(\mathbb{E}(g_q^{(0)}(X_i) g_r^{(0)}(X_j) | X_0) - \mathbb{E}(g_q^{(0)}(X_i) g_r^{(0)}(X_j)))\|_1. \end{aligned}$$

For three real-valued random variables  $A, B, C$ , define the numbers  $\bar{\alpha}(A, B)$  and  $\bar{\alpha}(A, B, C)$  by

$$\begin{aligned} \bar{\alpha}(A, B) &= \sup_{s, t \in \mathbb{R}} |\text{Cov}(\mathbf{1}_{A \leq s}, \mathbf{1}_{B \leq t})| \\ \bar{\alpha}(A, B, C) &= \sup_{s, t, u \in \mathbb{R}} |\mathbb{E}((\mathbf{1}_{A \leq s} - \mathbb{P}(A \leq s))(\mathbf{1}_{B \leq t} - \mathbb{P}(B \leq t))(\mathbf{1}_{C \leq u} - \mathbb{P}(C \leq u)))| \end{aligned}$$

(note that  $\bar{\alpha}(A, B, B) \leq \bar{\alpha}(A, B)$ ). Let

$$A = |g_l^{(0)}(X_0)| \text{sign}\{\mathbb{E}(g_q^{(0)}(X_i) g_r^{(0)}(X_j) | X_0) - \mathbb{E}(g_q^{(0)}(X_i) g_r^{(0)}(X_j))\},$$

and note that  $Q_A = Q_{g_l^{(0)}(X_0)}$ . From Proposition 6.1 and Lemma 6.1 in Dedecker and Rio (2007), we have that

$$\begin{aligned} \|g_l^{(0)}(X_0)(\mathbb{E}(g_q^{(0)}(X_i)g_r^{(0)}(X_j)|X_0) - \mathbb{E}(g_q^{(0)}(X_i)g_r^{(0)}(X_j)))\|_1 &= \mathbb{E}((A - \mathbb{E}(A))g_q^{(0)}(X_i)g_r^{(0)}(X_j)) \\ &\leq 16 \int_0^{\bar{\alpha}(A, g_q(X_i), g_r(X_j))/2} Q_{g_l^{(0)}(X_0)}(u)Q_{g_q(X_0)}(u)Q_{g_r(X_0)}(u)du. \end{aligned}$$

Note that  $Q_{g_l^{(0)}(X_0)} \leq Q_{g_l(X_0)} + \|g_l(X_0)\|_1$ . Hence, by Fréchet's inequality (1957),

$$\begin{aligned} \int_0^{\bar{\alpha}(A, g_q(X_i), g_r(X_j))/2} Q_{g_l^{(0)}(X_0)}(u)Q_{g_q(X_0)}(u)Q_{g_r(X_0)}(u)du \\ \leq 2 \int_0^{\bar{\alpha}(A, g_q(X_i), g_r(X_j))/2} Q_{g_l(X_0)}(u)Q_{g_q(X_0)}(u)Q_{g_r(X_0)}(u)du. \end{aligned}$$

Since  $\{g_i(x) \leq t\}$  is some interval of  $\mathbb{R}$ , we have that for  $j > i \geq 1$

$$\bar{\alpha}(A, g_q(X_i), g_r(X_j)) \leq 4\bar{\alpha}(A, X_i, X_j) \leq 4\alpha_2(i),$$

and for  $i = j$ ,

$$\bar{\alpha}(A, g_q(X_i), g_r(X_i)) \leq 4\bar{\alpha}(A, X_i, X_i) \leq 4\bar{\alpha}(X_0, X_i) \leq 4\alpha_1(i) \leq 4\alpha_2(i).$$

Since  $Q_{g_i(X_0)}(u) \leq Mu^{-1/p}$ , it follows that, for  $1 \leq i \leq j$ ,

$$\|g_l(X_0)(\mathbb{E}(g_q(X_i)g_r(X_j)|X_0) - \mathbb{E}(g_q(X_i)g_r(X_j)))\|_1 \leq \frac{32M^3p}{p-3}(2\alpha_2(i))^{\frac{p-3}{p}}.$$

Consequently, for any  $f$  in  $\mathcal{C}(M, p, \mu)$  with  $p > 3$ ,

$$\|Y_0(\mathbb{E}(Y_i Y_j | X_0) - \mathbb{E}(Y_i Y_j))\|_1 \leq \frac{32M^3p}{p-3}(2\alpha_2(i))^{\frac{p-3}{p}}.$$

In the same way,

$$2\sigma\|\mathbb{E}(Y_i Y_j | X_0) - \mathbb{E}(Y_i Y_j)\|_1 \leq \frac{32\sigma M^2 p}{p-2}(2\alpha_2(i))^{\frac{p-2}{p}}.$$

It follows that, for any  $1 \leq i \leq j$ ,

$$(5.6) \quad \|(|Y_0| + 2\sigma)(\mathbb{E}(Y_i Y_j | X_0) - \mathbb{E}(Y_i Y_j))\|_1 \leq D(M, p, \sigma)(\alpha_2(i))^{\frac{p-3}{p}}.$$

Combining (5.5) and (5.6), we infer that

$$\sum_{i=1}^m \sum_{j=i}^m \|(|Y_0| + 2\sigma)(\mathbb{E}(Y_i Y_j | X_0) - \mathbb{E}(Y_i Y_j))\|_1 = O(m^{1-\delta}),$$

and (5.4) easily follows. This completes the proof.

## 6. MOMENT INEQUALITIES

**Theorem 6.1.** *Let  $\mathbf{X} = (X_i)_{i \geq 0}$  be a stationary Markov chain with invariant measure  $\mu$  and transition kernel  $K$ . If  $f$  belong to  $\mathcal{C}(M, p, \mu)$  for some  $M > 0$  and some  $p > 2$ , then, for any  $2 \leq q < p$*

$$\left\| \sum_{i=1}^n (f(X_i) - \mu(f)) \right\|_q \leq \sqrt{2q} \left( n \|f(X_0) - \mu(f)\|_q^2 + 4M^2 \left( \frac{p}{p-q} \right)^{\frac{2}{q}} \sum_{k=1}^{n-1} (n-k) (2\alpha_1(k))^{\frac{2(p-q)}{pq}} \right)^{\frac{1}{2}}.$$

**Corollary 6.1.** *Let  $0 < \gamma < 1$ . Let  $f$  belong to  $\mathcal{C}(M, p, \mu)$  for some  $M > 0$  and some  $p > 2$ , and let  $2 \leq q < p$ .*

- (1) *If  $\gamma < 2(p-q)/(2(p-q) + pq)$ , then  $\|S_n(f - \nu_\gamma(f))\|_q = O(\sqrt{n})$ .*
- (2) *If  $2(p-q)/(2(p-q) + pq) \leq \gamma < 1$ , then, for any  $\epsilon > 0$ ,*

$$\|S_n(f - \nu_\gamma(f))\|_q = O\left(n^{1+\epsilon - \frac{(1-\gamma)(p-q)}{\gamma pq}}\right).$$

**Remark 6.1.** *Assume that  $\gamma < (p-2)/(2p-2)$ . By Chebichev inequality applied with  $2 \leq q < 2p(1-\gamma)/(\gamma p + 2(1-\gamma))$ , we infer from Item (1) that for any  $\epsilon > 0$ ,*

$$\nu_\gamma\left(\frac{1}{n} |S_n(f - \nu_\gamma(f))| > x\right) \leq \frac{C}{(nx^2)^{p(1-\gamma)/(\gamma p + 2(1-\gamma)) - \epsilon}}.$$

*Assume now that  $(p-2)/(2p-2) \leq \gamma < 1$ . By Chebichev inequality applied with  $q = 2$ , we infer from Item (2) that for any  $\epsilon > 0$ ,*

$$\nu_\gamma\left(\frac{1}{n} |S_n(f - \nu_\gamma(f))| > x\right) \leq \frac{C}{x^2 n^{(p-2)(1-\gamma)/\gamma p - \epsilon}}.$$

*When  $f$  is BV (case  $p = \infty$ ) and  $\gamma < 1$ , we obtain that, for any  $\epsilon > 0$  and any  $x > 0$ ,*

$$\nu_\gamma\left(\frac{1}{n} |S_n(f - \nu_\gamma(f))| > x\right) \leq \frac{C(x)}{n^{(1-\gamma)/\gamma - \epsilon}}.$$

*Note that Melbourne and Nicol (2007) obtained the same bound when  $f$  is  $\alpha$ -Hölder and  $\gamma < 1/2$ .*

**Two simple examples (continued).**

- (1) *Assume that  $f$  is positive and non increasing on  $[0, 1]$ , with  $f(x) \leq Cx^{-a}$  for some  $a > 0$ . If  $a < \frac{1}{2} - \gamma$  and  $2 \leq q < \frac{2(1-\gamma)}{\gamma+2a}$ , then  $\|S_n(f - \nu_\gamma(f))\|_q = O(\sqrt{n})$ . If now  $a < \frac{1-\gamma}{2}$  and  $2 \vee \frac{2(1-\gamma)}{\gamma+2a} \leq q < \frac{1-\gamma}{a}$ , then, for any  $\epsilon > 0$ ,*

$$\|S_n(f - \nu_\gamma(f))\|_q = O\left(n^{1+\epsilon - \frac{(1-\gamma-aq)}{\gamma q}}\right).$$

- (2) *Assume that  $f$  is positive and non increasing on  $[0, 1]$ , with  $f(x) \leq C(1-x)^{-a}$  for some  $a \geq 0$ . If  $a < \frac{1-2\gamma}{2(1-\gamma)}$  and  $2 \leq q < \frac{2(1-\gamma)}{\gamma+(1-\gamma)2a}$ , then  $\|S_n(f - \nu_\gamma(f))\|_q = O(\sqrt{n})$ . If  $a < \frac{1}{2}$  and  $2 \vee \frac{2(1-\gamma)}{\gamma+(1-\gamma)2a} \leq q < \frac{1}{a}$ , then, for any  $\epsilon > 0$ ,*

$$\|S_n(f - \nu_\gamma(f))\|_q = O\left(n^{1+\epsilon - \frac{(1-\gamma)(1-aq)}{\gamma q}}\right).$$

**Proof of Theorem 6.1.** From Proposition 4 in Dedecker and Doukhan (2003) (see also Theorem 2.5 in Rio (2000)), we have that, for any  $q \geq 2$ ,

$$\left\| \sum_{i=1}^n (f(X_i) - \mu(f)) \right\|_q \leq \sqrt{2q} \left( n \|f(X_0) - \mu(f)\|_q^2 + \sum_{k=1}^{n-1} (n-k) \| (f(X_0) - \mu(f)) (\mathbb{E}(f(X_k)|X_0) - \mu(f)) \|_{\frac{q}{2}} \right)^{\frac{1}{2}}.$$

Assume first that  $f = \sum_{i=1}^k a_i g_i$ , where  $\sum_{i=1}^k |a_i| \leq 1$ , and  $g_i$  belongs to  $\text{Mon}(M, p, \mu)$ . Clearly

$$\| (f(X_0) - \mu(f)) (\mathbb{E}(f(X_n)|X_0) - \mu(f)) \|_{q/2} \leq \sum_{i=1}^k \sum_{j=1}^k |a_i a_j| \| (g_i(X_0) - \mu(g_i)) (\mathbb{E}(g_j(X_n)|X_0) - \mu(g_j)) \|_{q/2}.$$

Applying Lemma 4.1, we obtain that

$$\| (f(X_0) - \mu(f)) (\mathbb{E}(f(X_n)|X_0) - \mu(f)) \|_{q/2} \leq 4M^2 \left( \frac{p}{p-q} \right)^{2/q} (2\alpha_1(n))^{\frac{2(p-q)}{pq}}.$$

Clearly, this inequality remains valid for any  $f$  in  $\mathcal{C}(M, p, \mu)$ , and the result follows.

## 7. THE EMPIRICAL DISTRIBUTION FUNCTION

**Theorem 7.1.** *Let  $\mathbf{X} = (X_i)_{i \geq 0}$  be a stationary Markov chain with invariant measure  $\mu$  and transition kernel  $K$ . Let  $F_n(t) = n^{-1} \sum_{i=1}^n \mathbf{1}_{X_i \leq t}$  and  $F_\mu(t) = \mu(\cdot - \infty, t]$ .*

- (1) *If  $\mathbf{X}$  is ergodic (in the ergodic theoretic sense) and if  $\sum_{k>0} \beta_1(k) < \infty$ , then, for any probability  $\pi$  on  $\mathbb{R}$ , the process  $\{\sqrt{n}(F_n(t) - F_\mu(t)), t \in \mathbb{R}\}$  converges in distribution in  $\mathbb{L}^2(\pi)$  to a tight Gaussian process  $G$  with covariance function*

$$\text{Cov}(G(s), G(t)) = C_{\mu, K}(s, t) = \mu(f_t^{(0)} f_s^{(0)}) + \sum_{k>0} \mu(f_t^{(0)} K^k f_s^{(0)}) + \sum_{k>0} \mu(f_s^{(0)} K^k f_t^{(0)}).$$

- (2) *Let  $(D(\mathbb{R}), d)$  be the space of cadlag functions equipped with the Skorohod metric  $d$ . If  $\beta_2(k) = O(k^{-2-\epsilon})$  for some  $\epsilon > 0$ , then the process  $\{\sqrt{n}(F_n(t) - F_\mu(t)), t \in \mathbb{R}\}$  converges in distribution in  $(D(\mathbb{R}), d)$  to a tight Gaussian process  $G$  with covariance function  $C_{\mu, K}$ .*

**Corollary 7.1.** *Let  $F_{n,\gamma}(t) = n^{-1} \sum_{i=1}^n \mathbf{1}_{T_\gamma^i \leq t}$ .*

- (1) *If  $0 < \gamma < 1/2$ , then, for any probability  $\pi$  on  $[0, 1]$ , the process  $\{\sqrt{n}(F_{n,\gamma}(t) - F_{\nu_\gamma}(t)), t \in [0, 1]\}$  converges in distribution in  $\mathbb{L}^2(\pi)$  to a tight Gaussian process  $G_\gamma$  with covariance function  $C_{\nu_\gamma, K_\gamma}$ .*
- (2) *If  $0 < \gamma < 1/3$ , the process  $\{\sqrt{n}(F_{n,\gamma}(t) - F_{\nu_\gamma}(t)), t \in [0, 1]\}$  converges in distribution in  $(D([0, 1]), d)$  to a tight Gaussian process  $G_\gamma$  with covariance function  $C_{\nu_\gamma, K_\gamma}$ .*

**Remark 7.1.** *Denote by  $\|\cdot\|_{p,\pi}$  the  $\mathbb{L}^p(\pi)$ -norm. If  $\gamma < 1/2$ , we have that, for any  $1 \leq p \leq 2$ ,*

$$(7.7) \quad \sqrt{n} \|F_{n,\gamma} - F_{\nu_\gamma}\|_{p,\pi} \quad \text{converges in distribution to} \quad \|G_\gamma\|_{p,\pi}.$$

*In particular, if  $\pi = \lambda$  is the Lebesgue measure on  $[0, 1]$  and  $q = p/(p-1)$ , we obtain that*

$$\frac{1}{\sqrt{n}} \sup_{\|f'\|_q \leq 1} |S_n(f - \nu_\gamma(f))| \quad \text{converges in distribution to} \quad \|G_\gamma\|_{p,\lambda}.$$

For  $p = 1$  and  $q = \infty$ , we obtain the limit distribution of the Kantorovič distance  $d_1(F_{n,\gamma}, F_{\nu_\gamma})$ :

$$\sqrt{nd_1}(F_{n,\gamma}, F_{\nu_\gamma}) = \frac{1}{\sqrt{n}} \sup_{f \in H_{1,1}} |S_n(f - \nu_\gamma(f))| \quad \text{converges in distribution to} \quad \int_0^1 |G_\gamma(t)| dt.$$

Now if  $\gamma < 1/3$ , the limit in (7.7) holds for any  $p \geq 1$ .

Note that, for Harris recurrent Markov chains, Item (2) of Theorem 7.1 holds as soon as the sum of the  $\beta$ -mixing coefficients of the chain is finite. Hence, we conjecture that Item (2) of Corollary 7.1 remains true for  $\gamma < 1/2$ .

**Proof of Theorem 7.1.** Item (1) has been proved in Dedecker and Merlevède (2007, Theorem 2, Item 2) and Item (2) in Dedecker and Prieur (2007, Proposition 2).

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