# Strong approximation for additive functionals of geometrically ergodic Markov chains

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Luminy, 15-19 February 2016

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### Strong approximation in the iid setting (1)

- Assume that  $(X_i)_{i\geq 1}$  is a sequence of iid centered real-valued random variables with a finite second moment  $\sigma^2$  and define  $S_n = X_1 + X_2 + \cdots + X_n$
- The ASIP says that a sequence (Z<sub>i</sub>)<sub>i≥1</sub> of iid centered Gaussian variables may be constructed is such a way that

$$\sup_{1\leq k\leq n} |S_k - \sigma B_k| = o(b_n) \text{ almost surely,}$$

where  $b_n = (n \log \log n)^{1/2}$  (Strassen (1964)).

 When (X<sub>i</sub>)<sub>i≥1</sub> is assumed to be in addition in L<sup>p</sup> with p > 2, then we can obtain rates in the ASIP:

$$b_n = n^{1/p}$$

(see Major (1976) for  $p \in ]2, 3]$  and Komlós, Major and Tusnády for p > 3).

### Strong approximation in the iid setting (2)

When (X<sub>i</sub>)<sub>i≥1</sub> is assumed to have a finite moment generating function in a neighborhood of 0, then the famous Komlós-Major-Tusnády theorem (1975 and 1976) says that one can construct a standard Brownian motion (B<sub>t</sub>)<sub>t≥0</sub> in such a way that

$$\mathbb{P}\Big(\sup_{k \le n} |S_k - \sigma B_k| \ge x + c \log n\Big) \le a \exp(-bx) \qquad (1)$$

where *a*, *b* and *c* are positive constants depending only on the law of  $X_1$ .

• (1) implies in particular that

$$\sup_{1 \le k \le n} \left| S_k - \sigma B_k \right| = O(\log n) \text{ almost surely}$$

 It comes from the Erdös-Rényi law of large numbers (1970) that this result is unimprovable.

### Strong approximation in the multivariate iid setting

- Einmahl (1989) proved that we can obtain the rate O((log n)<sup>2</sup>) in the almost sure approximation of the partial sums of iid random vectors with finite moment generating function in a neighborhood of 0 by Gaussian partial sums.
- Zaitsev (1998) removed the extra logarithmic factor and obtained the KMT inequality in the case of iid random vectors.
- What about KMT type results in the dependent setting?
- For functions of an iid sequence, see the recent paper Berkes, Liu and Wu (2014): the rate is o(n<sup>1/p</sup>).
- What about strong approximation in the Markov setting?

### What about strong approximation in the Markov setting?

- Let (ξ<sub>n</sub>) be an irreducible and aperiodic Harris recurrent Markov chain on a countably generated measurable state space (E, B). Let P(x, .) be the transition probability.
- We assume that the chain is positive recurrent. Let π be its (unique) invariant probability measure.
- Then there exists some positive integer m, some measurable function h with values in [0, 1] with π(h) > 0, and some probability measure ν on E, such that

$$P^m(x,A) \ge h(x)\nu(A)$$
.

- We assume that m = 1
- The Nummelin splitting technique (1984) allows to extend the Markov chain in such a way that the extended Markov chain has a recurrent atom. This allows regeneration.

### The Nummelin splitting technique

- Let  $Q(x, \cdot)$  be the sub-stochastic kernel defined by  $Q = P h \otimes \nu$
- The minorization condition allows to define an extended chain  $(\bar{\xi}_n, U_n)$  in  $E \times [0, 1]$  as follows.
- At time 0, U<sub>0</sub> is independent of ξ
  <sub>0</sub> and has the uniform distribution over [0, 1]; for any n ∈ N,

$$\mathbb{P}(\bar{\xi}_{n+1} \in A \mid \bar{\xi}_n = x, U_n = y) = \mathbf{1}_{y \le h(x)} \nu(A) + \mathbf{1}_{y > h(x)} \frac{Q(x, A)}{1 - h(x)}$$
$$:= \bar{P}((x, y), A)$$

and  $U_{n+1}$  is independent of  $(\bar{\xi}_{n+1}, \bar{\xi}_n, U_n)$  and has the uniform distribution over [0, 1].

P̃ = P̄ ⊗ λ (λ is the Lebesgue measure on [0, 1]) and (ξ̃<sub>n</sub>, U<sub>n</sub>) is an irreducible and aperiodic Harris recurrent chain, with unique invariant probability measure π ⊗ λ. Moreover (ξ̃<sub>n</sub>) is an homogenous Markov chain with transition probability P(x, .).

### Regeneration

• Define now the set C in  $E \times [0, 1]$  by

$$C = \{(x, y) \in E \times [0, 1] \text{ such that } y \leq h(x)\}.$$

For any (x, y) in C,  $\mathbb{P}(\bar{\xi}_{n+1} \in A \mid \bar{\xi}_n = x, U_n = y) = \nu(A)$ . Since  $\pi \otimes \lambda(C) = \pi(h) > 0$ , the set C is an atom of the extended chain, and it can be proven that this atom is recurrent.

Let

 $T_0 = \inf\{n \ge 1 : U_n \le h(\bar{\xi}_n)\}$  and  $T_k = \inf\{n > T_{k-1} : U_n \le h(\bar{\xi}_n)\}$ ,

and the return times  $(\tau_k)_{k>0}$  by  $\tau_k = T_k - T_{k-1}$ . Note that  $T_0$  is a.s. finite and the return times  $\tau_k$  are iid and integrable.

- Let  $S_n(f) = \sum_{k=1}^n f(\overline{\xi}_k)$ .
- The random vectors  $(\tau_k, S_{T_k}(f) S_{T_{k-1}}(f))_{k>0}$  are iid and their common law is the law of  $(T_0, S_{T_0}(f))$  under  $\mathbb{P}_C$ .

• Csáki and Csörgö (1995): If the r.v's  $S_{T_k}(|f|) - S_{T_{k-1}}(|f|)$  have a finite moment of order p for some p in ]2, 4] and if  $\mathbf{E}(\tau_k^{p/2}) < \infty$ , then one can construct a standard Wiener process  $(W_t)_{t\geq 0}$  such that

$$S_n(f) - n\pi(f) - \sigma(f) W_n = O(a_n)$$
 a.s. .

with  $a_n = n^{1/p} (\log n)^{1/2} (\log \log n)^{\alpha}$  and  $\sigma^2(f) = \lim_n \frac{1}{n} \operatorname{Var} S_n(f)$ .

- The above result holds for any bounded function *f* only if the return times have a finite moment of order *p*.
- The proof is based on the regeneration properties of the chain and on an application of the results of KMT (1975) to the partial sums of the iid random variables  $S_{T_{k+1}}(f) S_{T_k}(f)$ , k > 0.

On the proof of Csáki and Csörgö

- For any  $i \geq 1$ , let  $X_i = \sum_{\ell=T_{i-1}+1}^{T_i} f(\overline{\xi}_\ell)$ .
- Since the  $(X_i)_{i>0}$  are iid, if  $\mathbb{E}|X_1|^{2+\delta} < \infty$ , there exists a standard Brownian motion  $(W(t))_{t>0}$  such that

$$\sup_{k\leq n} \left| \sum_{i=1}^{k} X_i - \sigma(f) W(k) \right| = o(n^{1/(2+\delta)}) \quad a.s.$$

• Let  $\rho(n) = \max\{k : T_k \le n\}$ . If  $\mathbf{E}|\tau_1|^q < \infty$  for some  $1 \le q \le 2$ , then

$$\rho(n) = \frac{n}{\mathsf{E}(\tau_1)} + O(n^{1/q} (\log \log n)^{\alpha}) \quad a.s.$$

• 
$$\sum_{i=1}^{\rho(n)} X_i - W(\rho(n)) = o(n^{\frac{1}{2+\delta}})$$
,  $\sum_{i=1}^{\rho(n)} X_i - S_n(f) = o(n^{\frac{1}{2+\delta}})$  a.s.

• 
$$W(\rho(n)) - W(\frac{n}{\mathbf{E}(\tau_1)}) = O(n^{1/(2q)}(\log n)^{1/2}(\log \log n)^{\alpha}))$$
 a.s.

• With this method, no way to do better than  $O(n^{1/(2q)}(\log n)^{1/2})$  $(1 \le q \le 2)$  even if f is bounded and  $\tau_1$  has exponential moment.

## Link between the moments of return times and the coefficients of absolute regularity (1)

• For positive measures  $\mu$  and  $\nu$ , let  $\|\mu - \nu\|$  denote the total variation of  $\mu - \nu$ 

Set

$$\beta_n = \int_E \| P^n(x, .) - \pi \| d\pi(x) \, .$$

The coefficients  $\beta_n$  are called absolute regularity (or  $\beta$ -mixing) coefficients of the chain.

• Bolthausen (1980-1982): for any *p* > 1,

$$\mathsf{E}(\tau_1^p) = \mathsf{E}_C(T_0^p) < \infty \text{ if and only if } \sum_{n>0} n^{p-2}\beta_n < \infty.$$

## Link between the moments of return times and the coefficients of absolute regularity (2)

- Requiring β<sub>n</sub> = O(ρ<sup>n</sup>) for some real ρ with 0 < ρ < 1 is equivalent to say that the Markov chain is geometrically ergodic (see Nummelin and Tuominen (1982)).
- $\bullet\,$  If the Markov chain is GE then there exists a positive real  $\delta\,$  such that

$$\mathsf{E}ig(e^{t au_1}ig) < \infty$$
 and  $\mathsf{E}_{\pi}ig(e^{t au_0}ig) < \infty$  for any  $|t| \leq \delta$  .

• Heuristic: due to the decomposition in iid cycles, if f is bounded and the chain is geometrically ergodic, we can expect the same result as in the iid case when we have r.v.'s with exponential moments.

#### Main result: M. Rio (2015)

Assume that

$$eta_n = O(
ho^n) \;\; ext{for some real } 
ho \; ext{with } 0 < 
ho < 1,$$

 If f is bounded and such that π(f) = 0 then there exists a standard Wiener process (W<sub>t</sub>)<sub>t≥0</sub> and positive constants a, b and c depending on f and on the transition probability P(x, ·) such that, for any positive real x and any integer n ≥ 2,

$$\mathbb{P}_{\pi}\left(\sup_{k\leq n} \left|S_{k}(f) - \sigma(f)W_{k}\right| \geq c \log n + x\right) \leq a \exp(-bx).$$

where  $\sigma^{2}(f) = \pi(f^{2}) + 2\sum_{n>0} \pi(fP^{n}f) > 0.$ 

• Therefore  $\sup_{k \le n} |S_k(g) - \sigma(f)W_k| = O(\log n)$  a.s.

### A remark

• Let  $\mu$  be any law on E such that

$$\int_E \|\mathcal{P}^n(x,.) - \pi\| d\mu(x) = O(r^n) \text{ for some } r < 1.$$

Then  $\mathbb{P}_{\mu}(T_0 > n)$  decreases exponentially fast (see Nummelin and Tuominen (1982)).

 The result extends to the Markov chain (ξ<sub>n</sub>) with transition probability P and initial law μ. Some insights for the proof

- Let  $S_n(f) = \sum_{\ell=1}^n f(\bar{\xi}_\ell)$  and  $X_i = \sum_{\ell=T_{i-1}+1}^{T_i} f(\bar{\xi}_\ell)$ . Recall that  $(X_i, \tau_i)_{i>0}$  are iid.
- Let  $\alpha$  be the unique real such that  $Cov(X_k \alpha \tau_k, \tau_k) = 0$
- The random vectors (X<sub>i</sub> ατ<sub>i</sub>, τ<sub>i</sub>)<sub>i>0</sub> of ℝ<sup>2</sup> are then iid and their marginals are non correlated.
- By the multidimensional strong approximation theorem of Zaitsev (1998), there exist two independent standard Brownian motions  $(B_t)_t$  and  $(\widetilde{B}_t)_t$  such that

$$S_{T_n}(f) - \alpha(T_n - n\mathbf{E}(\tau_1)) - vB_n = O(\log n) \text{ a.s. } (1)$$

and

$$T_n - n\mathbf{E}(\tau_1) - \tilde{v}\widetilde{B}_n = O(\log n) \text{ a.s. } (2)$$

where  $v^2 = Var(X_1 - \alpha \tau_1)$  and  $\tilde{v}^2 = Var(\tau_1)$ .

• We associate to  $T_n$  a Poisson Process via (2).

• Let  $\lambda = \frac{(\mathbb{E}(\tau_1))^2}{\operatorname{Var}(\tau_1)}$ . Via KMT, one can construct a Poisson process N (depending on  $\widetilde{B}$ ) with parameter  $\lambda$  in such a way that

$$\gamma N(n) - n \mathbb{E}(\tau_1) - \tilde{v} \widetilde{B}_n = O(\log n)$$
 a.s.

• Therefore, via (2),

$$T_n - \gamma N(n) = O(\log n)$$
 a.s.

and then, via (1),

$$S_{\gamma N(n)}(f) - \alpha \gamma N(n) + \alpha n \mathbb{E}(\tau_1) - \nu B_n = O(\log n) \text{ a.s. } (3)$$

• The processes  $(B_t)_t$  and  $(N_t)_t$  appearing here are independent.

• Via (3), setting  $N^{-1}(k) = \inf\{t > 0 : N(t) \ge k\} := \sum_{\ell=1}^{k} \mathcal{E}_{\ell}$ ,

$$S_n(f) = vB_{N^{-1}(n/\gamma)} + \alpha n - \alpha \mathbb{E}(\tau_1)N^{-1}(n/\gamma) + O(\log n)$$
 a.s.

• If v = 0, the proof is finished. Indeed, by KMT, there exists a Brownian motion  $\widetilde{W}_n$  (depending on N) such that

$$\alpha n - \alpha \mathbb{E}(\tau_1) N^{-1}(n/\gamma) = \widetilde{W}_n + O(\log n)$$
 a.s.

• If  $v \neq 0$  and  $\alpha = 0$ , we have

$$S_n(f) = vB_{N^{-1}(n/\gamma)} + O(\log n) \text{ a.s.}$$

 Using Csörgö, Deheuvels and Horváth (1987) (B and N are independent), one can construct a Brownian motion W (depending on N) such that

$$B_{N^{-1}(n/\gamma)} - W_n = O(\log n)$$
 a.s.  $(*)$ ,

which leads to the expected result when  $\alpha = 0$ .

• However, in the case  $\alpha \neq 0$  and  $\nu \neq 0$ , we still have

$$S_n(f) = vB_{N^{-1}(n/\gamma)} + \alpha n - \alpha \mathbb{E}(\tau_1)N^{-1}(n/\gamma) + O(\log n) \text{ a.s.}$$

and then

$$S_n(f) = vW_n + \widetilde{W}_n + O(\log n)$$
 a.s.

• Since W and  $\widetilde{W}$  are not independent, we cannot conclude. Can we construct  $W_n$  independent of N (and then of  $\widetilde{W}$ ) such that (\*) still holds?

### The key lemma

- Let (B<sub>t</sub>)<sub>t≥0</sub> be a standard Brownian motion on the line and {N(t) : t ≥ 0} be a Poisson process with parameter λ > 0, independent of (B<sub>t</sub>)<sub>t≥0</sub>.
- Then one can construct a standard Brownian process  $(W_t)_{t\geq 0}$ independent of  $N(\cdot)$  and such that, for any integer  $n\geq 2$  and any positive real x,

$$\mathbb{P}\Big(\sup_{k\leq n} |B_k - \frac{1}{\sqrt{\lambda}} W_{N(k)}| \geq C \log n + x\Big) \leq A \exp(-Bx),$$

where A, B and C are positive constants depending only on  $\lambda$ .

 (W<sub>t</sub>)<sub>t≥0</sub> may be constructed from the processes (B<sub>t</sub>)<sub>t≥0</sub>, N(·) and some auxiliary atomless random variable δ independent of the σ-field generated by the processes (B<sub>t</sub>)<sub>t>0</sub> and N(·).

### Construction of W(1/3)

• It will be constructed from *B* by writing *B* on the Haar basis.

• For 
$$j \in \mathbb{Z}$$
 and  $k \in \mathbb{N}$ , let

$$e_{j,k} = 2^{-j/2} \left( \mathbf{1}_{]k2^{j},(k+\frac{1}{2})2^{j}]} - \mathbf{1}_{](k+\frac{1}{2})2^{j},(k+1)2^{j}]} \right),$$

and

$$Y_{j,k} = \int_0^\infty e_{j,k}(t) dB(t) = 2^{-j/2} \left( 2B_{(k+\frac{1}{2})2^j} - B_{k2^j} - B_{(k+1)2^j} \right).$$

Then, since  $(e_{j,k})_{j\in\mathbb{Z},k\geq 0}$  is a total orthonormal system of  $\ell^2(\mathbb{R})$ , for any  $t\in\mathbb{R}^+$ ,

$$B_t = \sum_{j \in \mathbb{Z}} \sum_{k \ge 0} \left( \int_0^t e_{j,k}(t) dt \right) Y_{j,k}.$$

• To construct W, we modify the  $e_{j,k}$ .

## Construction of W(2/3)

- Let  $E_j = \{k \in \mathbb{N} : N(k2^j) < N((k+\frac{1}{2})2^j) < N((k+1)2^j)\}$
- For  $j \in \mathbb{Z}$  and  $k \in E_j$ , let

$$f_{j,k} = c_{j,k}^{-1/2} \left( b_{j,k} \mathbf{1}_{]N(k2^{j}),N((k+\frac{1}{2})2^{j})]} - a_{j,k} \mathbf{1}_{]N((k+\frac{1}{2})2^{j}),N((k+1)2^{j})]} \right),$$

where

$$a_{j,k} = N((k+\frac{1}{2})2^j) - N(k2^j)$$
,  $b_{j,k} = N((k+1)2^j) - N((k+\frac{1}{2})2^j)$ ,

and  $c_{j,k} = a_{j,k} b_{j,k} (a_{j,k} + b_{j,k})$ 

- $(f_{j,k})_{j \in \mathbb{Z}, k \in E_j}$  is an orthonormal system whose closure contains the vectors  $\mathbf{1}_{]0,N(t)]}$  for  $t \in \mathbb{R}^+$  and then the vectors  $\mathbf{1}_{]0,\ell]}$  for  $\ell \in \mathbb{N}^*$ .
- Setting  $f_{j,k} = 0$  if  $k \notin E_j$ , we define

$$W_\ell = \sum_{j \in \mathbb{Z}} \sum_{k \ge 0} \left( \int_0^\ell f_{j,k}(t) dt \right) Y_{j,k}$$
 for any  $\ell \in \mathbb{N}^*$  and  $W_0 = 0$ 

## Construction of W(3/3)

$$W_\ell = \sum_{j \in \mathbb{Z}} \sum_{k \ge 0} \left( \int_0^\ell f_{j,k}(t) dt \right) Y_{j,k}$$
 for any  $\ell \in \mathbb{N}^*$  and  $W_0 = 0$ 

- Conditionally to N,  $(f_{j,k})_{j \in \mathbb{Z}, k \in E_j}$  is an orthonormal system and  $(Y_{j,k})$  is a sequence of iid  $\mathcal{N}(0, 1)$ , independent of N.
- Hence, conditionally to N,  $(W_{\ell})_{\ell \geq 0}$  is a Gaussian sequence such that  $Cov(W_{\ell}, W_m) = \ell \wedge m$ .
- Therefore this Gaussian sequence is independent of N
- By the Skorohod embedding theorem, we can extend it to a standard Wiener process (W<sub>t</sub>)<sub>t</sub> still independent of N.

Thank you for your attention!

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