

# The CLT for Markov chains with a countable state space embedded in the space $l_p$

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## Abstract

We find necessary and sufficient conditions for the CLT for Markov chains with a countable state space embedded in the space  $l_p$  for  $p \geq 1$ . This result is an extension of the uniform CLT over the family of indicator functions in Levental (Stochastic Processes Appl. 34 (1990)245-253), where the result is equivalent to our case  $p = 1$ . A similar extension for the uniform CLT over a family of possibly unbounded functions in Tsai(Taiwan. J. Math. 1(4)(1997)481-498) is also obtained.

## 1 Introduction and results

Let  $\{X_i\}_{i \geq 0}$  be a Markov chain with state space  $\mathbf{N} = \{1, 2, 3, \dots\}$ . We assume that the chain is irreducible and positive recurrent. It is well known that  $\{X_i\}$  has a unique invariant probability measure, which we denote by  $\pi$ . Let

$$N_1 = \min\{n \geq 1 : X_n = 1\}$$

and

$$N_i = \min\{n > N_{i-1} : X_n = 1\}$$

for  $i \geq 2$ . This random variable is called the  $i$ -th entrance time of the state 1. Let

$$m_{i,j} = E(\min\{n : n \geq 1, X_n = j \mid X_0 = i\})$$

denote the mean entrance time to state  $j$  for a Markov chain starting at state  $i$ .

Let  $f \in L_1(\mathbf{N}, \pi)$ . We denote the centered sum by

$$S_n(f) = \sum_{i=1}^n (f(X_i) - \pi(f)).$$

It is well known that  $S_n(f)/\sqrt{n}$  converges in law to a normal distribution if  $f$  is bounded and the Markov chain satisfies the condition (see section 16 in Chung (1967))

$$E(N_2 - N_1)^2 < \infty. \quad (1.1)$$

Levental (1990) extended this CLT to a uniform CLT. That is, we say that the CLT holds uniformly over  $\mathcal{F} \subset L_2(\mathbf{N}, \pi)$  if the process

$$\{S_n(f)/\sqrt{n}\}_{f \in \mathcal{F}}$$

converges weakly to a Gaussian process indexed by  $\mathcal{F}$  whose sample paths are uniformly continuous with respect to the  $L_2$ -norm metric of  $L_2(\mathbf{N}, \pi)$ . By assuming that the Markov chain satisfies condition (1.1), Levental proved that the CLT holds uniformly over  $\mathcal{F} = \{1_A : A \subseteq \mathbf{N}\}$  if and only if

$$\sum_{k=1}^{\infty} \pi(k) m_{1,k}^{\frac{1}{2}} < \infty. \quad (1.2)$$

Levental's result is a generalization of a result by Durst and Dudley (1981) for i.i.d. observations, where the condition is

$$\sum_{k=1}^{\infty} \pi^{\frac{1}{2}}(k) < \infty. \quad (1.3)$$

Of course, in the i.i.d. case  $m_{1,k} = (\pi(k))^{-1}$ , so (1.2) coincides with (1.3).

The uniform CLT with a countable space  $\mathbf{N}$  can be formulated as results in the space  $l_p$ . Here and throughout this paper,  $l_p$  denotes the space of real functions on  $\mathbf{N}$  for which

$$\|f\|_p = \left( \sum_i |f(i)|^p \right)^{\frac{1}{p}} < \infty,$$

( $\sup_i |f(i)| < \infty$  if  $p = \infty$ ) for all  $0 < p \leq \infty$ . Let  $B_p$  be the unit ball of  $l_p$ . Let  $\mathbf{N}$  be embedded into the canonical basis of  $l_p$ , that is, define  $\xi : \mathbf{N} \rightarrow l_p$  by  $\xi(k) = e_k$ , where

$$e_k = (0, 0, \dots, 0, 1^{k\text{-th}}, 0, \dots) \in l_1, \quad k \in \mathbf{N}$$

is the canonical basis of  $l_p$ . We say  $\xi(X_i)$  satisfies the CLT in the space  $l_p$  and denote by

$$\xi(X_i) \in CLT(l_p),$$

if and only if the normalized centered sum

$$\frac{1}{\sqrt{n}} \sum_{i=1}^n (\xi(X_i) - \xi(\pi))$$

converges weakly to a Gaussian measure in the space  $l_p$ . Note that

$$f(X_i) = \langle f, \xi(X_i) \rangle,$$

for all  $f$ . Thus  $\xi(X_i) \in CLT(l_1)$  if and only if the CLT holds uniformly over  $B_\infty$ , or, what is equivalent, over  $\{1_A : A \subseteq \mathbf{N}\}$ . In other words, in Levental (1990), the uniform CLT is equivalent to  $\xi(X_i) \in CLT(l_1)$ .

Our purpose is to find the condition for  $\xi(X_i) \in CLT(l_p)$  for  $p > 1$ , which is equivalent to that the CLT holds uniformly over  $B_q$ , where  $q = p/(p-1)$  is the conjugate exponent to  $p$ . We obtain necessary and sufficient conditions for  $\xi(X_i) \in CLT(l_p)$ ,  $1 \leq p < \infty$ , in the following theorem.

**Theorem 1.** For  $1 \leq p \leq 2$ ,  $\xi(X_i) \in CLT(l_p)$  if and only if (1.1) holds and

$$\sum_{k=1}^{\infty} \pi^p(k) m_{1,k}^{\frac{p}{2}} < \infty. \quad (1.4)$$

For  $2 < p < \infty$ ,  $\xi(X_i) \in CLT(l_p)$  if and only if (1.1) holds.

Note that condition (1.1) actually implies (1.4) for  $p > 2$ .

The proof of the result uses first blocking between the successive  $N_i$  to reduce the problem to independent random variables, and then applies the CLT for independent random variables taking values in  $l_p$ ; see Jain (1977) and Pisier and Zinn (1978).

The families of functions over which the CLT for Markov chains holds uniformly need not be bounded. In [7] we give a necessary and sufficient condition for the CLT to hold uniformly over collections of functions of the form  $\mathcal{F} = \{f : |f| \leq F\}$ , where  $F$  is a positive function in  $L_1(S, \pi)$ . This condition is

$$\sum_{k=1}^{\infty} F(k) \pi(k) m_{1,k}^{\frac{1}{2}} < \infty.$$

We derive a similar result in the space  $l_p$  that is relative to the uniform CLT over  $\mathcal{F} = \{f : |f| \leq F\}$ . Define  $\xi_F : \mathbf{N} \rightarrow l_p$  by

$$\xi_F(k) = F(k)e_k.$$

Then

$$f(X_i) = \langle f, \xi(X_i) \rangle = \left\langle \frac{f}{F}, \xi_F(X_i) \right\rangle,$$

for all  $f \in \{f : |f| \leq F\}$ . Thus  $\xi_F(X_i) \in CLT(l_1)$  if and only if the CLT holds uniformly over  $\mathcal{F} = \{f : |f| \leq F\}$ . We extend the result to  $p > 1$  as follows.

**Theorem 2.** For  $1 \leq p \leq 2$ ,  $\xi_F(X_i) \in CLT(l_p)$  if and only if (1.1) holds and

$$\sum_{k=1}^{\infty} F^p(k) \pi^p(k) m_{1,k}^{\frac{p}{2}} < \infty. \quad (1.5)$$

For  $2 < p < \infty$ ,  $\xi_F(X_i) \in CLT(l_p)$  if and only if (1.1) and (1.5) hold and

$$t^2 P \left( \left\| \sum_{N_1 < i \leq N_2} \xi_F(X_i) \right\|_p > t \right) \rightarrow 0 \text{ as } t \rightarrow \infty. \quad (1.6)$$

## 2 Proof

We will use the CLT for independent random variables taking values in  $l_p$ ; see Corollary 1 in Jain (1977) and Theorem 5.1 in Pisier and Zinn (1978).

Let  $\{Y_i\}$  be i.i.d. random variables in the space  $l_p$ .

**Theorem A.** (Jain, 1977)

For  $1 \leq p \leq 2$ ,  $Y_i \in CLT(l_p)$  if and only if

$$\sum_{k=1}^{\infty} \left( E \left[ Y_1^{(k)} \right]^2 \right)^{\frac{p}{2}} < \infty, \quad (2.1)$$

where  $Y_1^{(k)}$  is the  $k$ -th component of  $Y_1$ .

**Theorem B.** (Pisier and Zinn, 1978)

For  $2 < p < \infty$ ,  $Y_i \in CLT(l_p)$  if and only if (2.1) holds and

$$t^2 P(\|Y_1\|_p > t) \rightarrow 0 \text{ as } t \rightarrow \infty. \quad (2.2)$$

In particular, if  $E \|Y_1\|_p^2 < \infty$  then (2.1) and (2.2) hold for  $2 < p < \infty$ .

Let  $\{X_i\}_{i \geq 0}$  be a positive recurrent and irreducible Markov chain taking values in  $\mathbf{N}$  with the unique invariant probability measure  $\pi$ . Denote by  $N_i$  the  $i$ -th entrance time of state 1. Let

$$l(n) = \max\{i : N_i \leq n\}$$

be the number of times state 1 is visited. We also need the following lemma; see Tsai (1997).

**Lemma C.** The random variable  $(l(n) - \frac{n}{m_{1,1}})/\sqrt{n}$  converges in law to a normal distribution if and only if (1.1) holds.

The centered sum  $S_n$  can be divided into four parts

$$\sum_{1 \leq i \leq N_1} (\xi(X_i) - \xi(\pi)) + \sum_{k=1}^{\left\lfloor \frac{n}{m_{1,1}} \right\rfloor} Z_k + \left( \sum_{k=1}^{l(n)-1} Z_k - \sum_{k=1}^{\left\lfloor \frac{n}{m_{1,1}} \right\rfloor} Z_k \right) + \sum_{N_{l(n)} < i \leq n} (\xi(X_i) - \xi(\pi)),$$

where the blocks  $\{Z_k\}$ ,

$$Z_k = \sum_{N_k < i \leq N_{k+1}} (\xi(X_i) - \xi(\pi)),$$

are i.i.d.. Since the first term does not depend on  $n$

$$\lim_{n \rightarrow \infty} \frac{1}{\sqrt{n}} \left\| \sum_{1 \leq i \leq N_1} (\xi(X_i) - \xi(\pi)) \right\|_p \rightarrow 0 \quad a.s. \quad (2.3)$$

for all  $p \geq 1$ . Regarding the fourth part, it is bounded in probability, (see Theorem 8 in section 14 in Chung), thus

$$\lim_{n \rightarrow \infty} \frac{1}{\sqrt{n}} \left\| \sum_{N_{l(n)} < i \leq n} (\xi(X_i) - \xi(\pi)) \right\|_p \rightarrow 0 \text{ in probability} \quad (2.4)$$

for all  $p \geq 1$ . For the third part we need the following two propositions.

**Proposition 1.** Let  $Z_1, Z_2, \dots$  be i.i.d. random variables taking values in the space  $l_p$  with  $EZ_1 = 0$  and  $E \|Z_1\|_p < \infty$  for some  $p \geq 1$ . Assume that (1.1) holds. Then

$$\lim_{n \rightarrow \infty} \frac{1}{\sqrt{n}} \left\| \sum_{k=1}^{l(n)-1} Z_k - \sum_{k=1}^{\lfloor \frac{n}{m_{1,1}} \rfloor} Z_k \right\|_p \rightarrow 0 \text{ in probability.} \quad (2.5)$$

**Proof.** Fix  $\varepsilon > 0$  and claim that

$$\lim_{n \rightarrow \infty} P \left( \frac{1}{\sqrt{n}} \left\| \sum_{k=1}^{l(n)-1} Z_k - \sum_{k=1}^{\lfloor \frac{n}{m_{1,1}} \rfloor} Z_k \right\|_p > \varepsilon \right) = 0.$$

Given  $\delta > 0$ , by Lemma C there is a  $c > 0$  such that

$$P \left( \frac{n}{m_{1,1}} - c\sqrt{n} \leq l(n) - 1 \leq \frac{n}{m_{1,1}} + c\sqrt{n} \right) \geq 1 - \delta.$$

Observe that, on  $\{\frac{n}{m_{1,1}} - c\sqrt{n} \leq l(n) - 1 \leq \frac{n}{m_{1,1}} + c\sqrt{n}\}$ ,

$$\left\| \sum_{k=1}^{l(n)-1} Z_k - \sum_{k=1}^{\lfloor \frac{n}{m_{1,1}} \rfloor} Z_k \right\|_p \leq \max_{1 \leq i \leq c\sqrt{n}} \left\| \sum_{k=1}^i Z_k \right\|_p.$$

Then, taking  $m = \lfloor c\sqrt{n} \rfloor$ ,

$$\begin{aligned}
& P \left( \frac{1}{\sqrt{n}} \left\| \sum_{k=1}^{l(n)-1} Z_k - \sum_{k=1}^{\lfloor \frac{n}{m_{1,1}} \rfloor} Z_k \right\|_p > \varepsilon \right) \\
& \leq \delta + P \left( \frac{1}{\sqrt{n}} \max_{1 \leq i \leq c\sqrt{n}} \left\| \sum_{k=1}^i Z_k \right\|_p > \varepsilon \right) \\
& \leq \delta + P \left( \frac{1}{m} \max_{1 \leq i \leq m} \left\| \sum_{k=1}^i Z_k \right\|_p > \frac{\varepsilon}{c} \right).
\end{aligned}$$

Since  $EZ_1 = 0$  and  $E\|Z_1\|_p < \infty$ , the strong law of large numbers in the space  $l_p$  (see Corollary 7.10 in Ledoux and Talagrand (1991)) gives

$$\frac{1}{m} \left\| \sum_{k=1}^m Z_k \right\|_p \rightarrow 0 \quad a.s. \text{ as } m \rightarrow \infty,$$

thus

$$\frac{1}{m} \max_{1 \leq i \leq m} \left\| \sum_{k=1}^i Z_k \right\|_p \rightarrow 0 \quad a.s. \text{ as } m \rightarrow \infty.$$

Consequently,

$$P \left( \frac{1}{m} \max_{1 \leq i \leq m} \left\| \sum_{k=1}^i Z_k \right\|_p > \frac{\varepsilon}{c} \right) \rightarrow 0 \text{ as } m \rightarrow \infty.$$

Since  $\delta$  is arbitrary we have

$$P \left( \frac{1}{\sqrt{n}} \left\| \sum_{k=1}^{l(n)-1} Z_k - \sum_{k=1}^{\lfloor \frac{n}{m_{1,1}} \rfloor} Z_k \right\|_p > \varepsilon \right) \rightarrow 0.$$

**Proposition 2.** If (1.1) holds then  $E[(\|Z_1\|_p)^2] < \infty$  for all  $p \geq 1$ .

**Proof.** Let  $\text{card}\{A\}$  denote the cardinality of  $A$ . We have

$$\begin{aligned} & \left\| \sum_{N_1 < i \leq N_2} \xi(X_i) \right\|_p \\ &= \left( \sum_{k=1}^{\infty} (\text{card}\{i : X_i = k, N_1 < i \leq N_2\})^p \right)^{\frac{1}{p}} \\ &\leq \left( \left( \sum_{k=1}^{\infty} \text{card}\{i : X_i = k, N_1 < i \leq N_2\} \right)^p \right)^{\frac{1}{p}} \\ &= N_2 - N_1. \end{aligned}$$

Thus

$$E \left[ (\|Z_1\|_p)^2 \right] \leq E(N_2 - N_1)^2 < \infty.$$

**Proof of Theorem 1.** First assume that (1.1) holds. For all  $p \geq 1$ , we obtain, by using (2.3), (2.4) and (2.5),

$$\left\| \frac{1}{\sqrt{n}} \left( S_n - \sum_{k=1}^{\lfloor \frac{n}{m_{1,1}} \rfloor} Z_k \right) \right\|_p \rightarrow 0 \text{ in probability.}$$

Thus  $\xi(X_i) \in CLT(l_p)$  if and only if  $Z_i \in CLT(l_p)$ .

For  $1 \leq p \leq 2$ , by Theorem A,  $Z_i \in CLT(l_p)$  if and only if

$$\sum_{k=1}^{\infty} \left( E \left[ Z_1^{(k)} \right]^2 \right)^{\frac{p}{2}} < \infty. \quad (2.6)$$

Note that

$$Z_1^{(k)} = \sum_{N_1 < i \leq N_2} (1_{\{X_i=k\}} - \pi(k)).$$

The second moment of a block (Theorem 7) in section 14 in Chung gives

$$E \left[ Z_1^{(k)} \right]^2 = 2m_{1,1}\pi^2(k) (m_{1,k} + m_{k,1}) - m_{1,1}\pi(k) - m_{1,1}^2\pi^2(k),$$

and

$$\sum_{k=1}^{\infty} \pi(k)m_{k,1} < \infty. \quad (2.7)$$

By (2.7)

$$2m_{k,1} \leq \frac{1}{\pi(k)} = m_{k,k} \leq m_{1,k} + m_{k,1},$$

and

$$m_{1,k} \leq m_{1,k} + m_{k,1} \leq 2m_{1,k},$$

except for finitely many  $k$ . Hence the condition (2.6) holds if and only if the condition (1.4) holds. Therefore we have that  $\xi(X_i) \in CLT(l_p)$  if and only if (1.4) holds for  $1 \leq p \leq 2$ .

For  $2 < p < \infty$ , we have  $E[(\|Z_1\|_p)^2] < \infty$  by Proposition 2. By Theorem B, we obtain that  $Z_i \in CLT(l_p)$ , which is proved to be equivalent to  $\xi(X_i) \in CLT(l_p)$ . Note that  $E[(\|Z_1\|_p)^2] < \infty$  implies (2.6), which is proved to be equivalent to the condition (1.4). Thus (1.1) implies (1.4) for  $2 < p < \infty$ .

It remains to prove that if  $\xi(X_i) \in CLT(l_p)$  for some  $1 \leq p < \infty$  then (1.1) holds. Note that  $\xi(X_i) \in CLT(l_p)$  implies

$$\frac{\sum_{i=1}^n 1_{\{1\}}(X_i) - n\pi(1)}{\sqrt{n}} \text{ converges in law to } \langle 1_{\{1\}}, G \rangle.$$

Since  $\sum_{i=1}^n 1_{\{1\}}(X_i) = l(n)$  and  $\pi(1) = \frac{1}{m_{1,1}}$ , the random variable

$$\frac{1}{\sqrt{n}} \left( l(n) - \frac{n}{m_{1,1}} \right)$$

converges in law to a normal distribution  $\langle 1_{\{1\}}, G \rangle$ . Hence we have (1.1) by Lemma C.

**Proof of Theorem 2.** Assume that (1.1) holds. Let

$$Z_k^F = \sum_{N_k < i \leq N_{k+1}} (\xi_F(X_i) - \xi_F(\pi)).$$

Then the assumption of Proposition 1 holds, that is we have

$$E \|Z_1^F\|_p \leq E \|Z_1^F\|_1 \leq E \left( \sum_{N_1 < i \leq N_2} F(X_i) \right) = E(N_2 - N_1)\pi(F) < \infty,$$

since, by assumption,  $F$  is a positive function in  $L_1(\mathbf{N}, \pi)$ . As in the proof of Theorem 1, we can obtain that  $\xi_F(X_i) \in CLT(l_p)$  if and only if  $Z_k^F \in CLT(l_p)$  for all  $p \geq 1$ .

For  $1 \leq p \leq 2$ , we have that  $Z_k^F \in CLT(l_p)$  if and only if

$$\sum_{k=1}^{\infty} \left( E \left[ (Z_1^F)^{(k)} \right]^2 \right)^{\frac{p}{2}} < \infty, \tag{2.8}$$

by Theorem A. Note that

$$\left( E \left[ (Z_1^F)^{(k)} \right]^2 \right)^{\frac{p}{2}} = F^p(k) \left( 2m_{1,1}\pi^2(k) (m_{1,k} + m_{k,1}) - m_{1,1}\pi(k) - m_{1,1}^2\pi^2(k) \right)^{\frac{p}{2}},$$

and thus (2.8) holds if and only if the condition (1.5) holds. Therefore we have that  $\xi_F(X_i) \in CLT(l_p)$  if and only if (1.5) holds.

For  $2 < p < \infty$ ,  $Z_k^F \in CLT(l_p)$  if and only if (2.8) holds and

$$t^2 P(\|Z_1^F\|_p > t) \rightarrow 0 \text{ as } t \rightarrow \infty, \quad (2.9)$$

by Theorem B. Note that (2.9) is equivalent to (1.6). Thus we have that  $\xi_F(X_i) \in CLT(l_p)$  if and only if (1.5) and (1.6) hold.

The proof of the necessity of the condition (1.1) is the same as the proof of Theorem 1.

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