

# On uniform closeness of local times of Markov chains and i.i.d. sequences

Serguei Popov  
University of Campinas

Joint work with D.F. de Bernardini and C. Gallesco

June 13, 2018

# Outline

- 1 Motivating example
- 2 Main results
- 3 Elements of proof of Theorem 1

# Outline

- 1 Motivating example
- 2 Main results
- 3 Elements of proof of Theorem 1

# Motivating example

Let  $X$  be a Markov chain on the state space  $\Sigma = \{0, 1\}$ , with transition probabilities:

$\mathbb{P}[X_{n+1} = k \mid X_n = k] = 1 - \mathbb{P}[X_{n+1} = 1 - k \mid X_n = k] = \frac{1}{2}(1 + \varepsilon)$   
for  $k = 0, 1$ , where  $\varepsilon > 0$  is **small**.

By symmetry,  $\pi = (\frac{1}{2}, \frac{1}{2})$  is the stationary distribution of this Markov chain.

Next, let  $Y$  be a sequence of i.i.d. Bernoulli random variables with success probability  $\frac{1}{2}$ .

# Introductory example

Consider in this case,

$$L_n^Z(0) = \sum_{j=1}^n \mathbf{1}_{\{Z_j=0\}}$$

for  $Z = X$  or  $Y$ . What can we say about  $d_{\text{TV}}(L_n^X(0), L_n^Y(0))$ ?

# Introductory example

Well, the random variable  $L_n^Y(0)$  has the Binomial distribution with parameters  $n$  and  $\frac{1}{2}$ , so it is approximately **Normal** with mean  $\frac{n}{2}$  and standard deviation  $\frac{\sqrt{n}}{2}$ .

As for  $L_n^X(0)$ , it is approximately **Normal** with mean  $\frac{n}{2}$  and standard deviation  $\sqrt{n}(\frac{1}{2} + O(\varepsilon))$ .

It is not difficult obtain that the total variation distance between these two Normals is  $O(\varepsilon)$ , **uniformly** in  $n$ .

This *suggests* that the total variation distance between  $L_n^X(0)$  and  $L_n^Y(0)$  should be also of order  $\varepsilon$  uniformly in  $n$ .

## Why bother with local times?

Many quantities of interest that can be expressed in terms of local times only, such as,

- hitting time of a site  $x$ :  $\tau(x) = \min\{n : L_n(x) > 0\}$ ;
- cover time:  $\min\{n : L_n(x) > 0 \text{ for all } x \in \Sigma\}$ ;
- blanket time :  $\min\{n \geq 1 : L_n(x) \geq \delta n \pi(x)\}$ , where  $\delta \in (0, 1)$ ;
- disconnection time;
- the set of favorite (most visited) sites;
- and so on...

# Outline

- 1 Motivating example
- 2 Main results**
- 3 Elements of proof of Theorem 1



# Main results

Let  $(\Sigma, d)$  be a compact metric space, with  $\mathcal{B}(\Sigma)$  representing its Borel  $\sigma$ -algebra.

## Assumption (A1)

*We assume that  $(\Sigma, d)$  is of polynomial class: there exist some  $\beta \geq 0$  and  $\varphi \geq 1$  such that for all  $r \in (0, 1]$ , the number of open balls of radius  $r$  needed to cover  $\Sigma$  is smaller than or equal to  $\varphi r^{-\beta}$ .*

Example:  $\Sigma$  finite, endowed with the discrete metric

$$d(x, y) = \mathbf{1}_{\{x \neq y\}}, \text{ for } x, y \in \Sigma.$$

We can choose  $\beta = 0$  and  $\varphi = |\Sigma|$ .

# Main results

Consider a Markov chain  $X = (X_i)_{i \geq 1}$  with transition kernel  $\mathfrak{P}(x, dy)$  on  $(\Sigma, \mathcal{B}(\Sigma))$ , and unique invariant probability measure  $\pi$  such that  $\mathfrak{P}(x, \cdot) \ll \pi(\cdot)$  for all  $x \in \Sigma$ .

Denote by  $p(x, \cdot)$  the *density* of  $\mathfrak{P}(x, \cdot)$  with respect to  $\pi$ : for  $x \in \Sigma$ ,

$$\mathfrak{P}(x, A) = \int_A p(x, y) \pi(dy), \text{ for all } A \in \mathcal{B}(\Sigma).$$

We also consider

## Assumption (A2)

*Assume that the density  $p(x, \cdot)$  is uniformly Hölder continuous, that is, there exist constants  $\kappa > 0$  and  $\gamma \in (0, 1]$  such that for all  $x, z, z' \in \Sigma$ ,*

$$|p(x, z) - p(x, z')| \leq \kappa d^\gamma(z, z').$$

# Main results

We assume that the chain  $X$  starts with some probability law absolutely continuous with respect to  $\pi$  and we denote by  $\nu$  its density.

## Assumption (A3)

*There exists  $\varepsilon \in (0, 1/2)$  such that*

$$\sup_{x,y \in \Sigma} |p(x,y) - 1| \vee \sup_{x \in \Sigma} |\nu(x) - 1| \leq \varepsilon.$$

# Main results

Let us denote also by  $Y = (Y_n)_{n \geq 1}$  a sequence of i.i.d. random variable with law  $\pi$ .

We have the following

## Theorem 1

Under Assumptions 1-3, there exists a universal constant  $K > 0$  such that, for all  $n \geq 1$ , it holds that

$$d_{\text{TV}}(L_n^X, L_n^Y) \leq K\varepsilon \sqrt{1 + \ln(\varphi 2^\beta) + \frac{\beta}{\gamma} \ln\left(\frac{\kappa \vee (2\varepsilon)}{\varepsilon}\right)}.$$

# Main results

**Case of a finite state space  $\Sigma$ , endowed with the discrete metric:**

Choosing  $\beta = 0$  and  $\varphi = |\Sigma|$ , Theorem 1 boils down to

$$d_{\text{TV}}(L_n^X, L_n^Y) \leq K\varepsilon\sqrt{1 + \ln |\Sigma|},$$

for all  $n \geq 1$ .

Recall motivating example!

# Main results

We also have

## Theorem 2

Under Assumptions 1-3, there exists a positive constant  $K' = K'(\beta, \varphi, \kappa, \gamma, \varepsilon)$ , such that, for all  $n \geq 1$ , it holds that

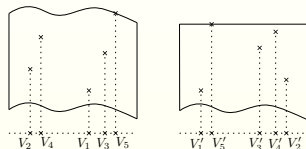
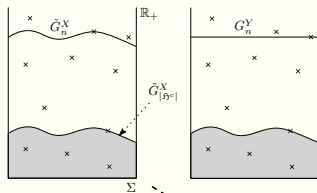
$$d_{\text{TV}}(L_n^X, L_n^Y) \leq 1 - K'.$$

# Outline

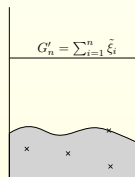
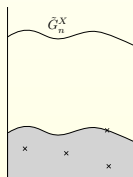
- 1 Motivating example
- 2 Main results
- 3 Elements of proof of Theorem 1**

# Idea of the proof: the soft local times method

naïve coupling  
does not work well



erase the points above the “dependent” part



use  $\tilde{\xi}_1, \dots, \tilde{\xi}_n$   
to construct  $\eta_Y$   
until level  $G'_n$

resample, using the maximal coupling



# Thanks!