

The Berry-Esseen Theorem for Functionals of Discrete Markov Chains

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Summary. The error bound $O(1/\sqrt{n})$ is derived in the central limit theorem for partial sums $\sum_{j=1}^n f(\xi_j)$ where ξ_j is a recurrent discrete Markov chain and f is a real valued function on the state space. In particular it is shown that for bounded f and starting distribution dominated by some multiple of the stationary one, it is sufficient for the chain to have recurrence times with third moments on order to get this bound.

§ 1. Introduction

Let I be an at most countable set of states, $(p_{ij})_{i,j \in I}$ a stochastic matrix (i.e. $p_{ij} \geq 0$, $\sum_{j \in I} p_{ij} = 1$ for all $i \in I$) and $X = (\Omega, \mathfrak{A}, \xi_n, P_i)$ a Markov chain with transition probabilities p_{ij} ; i.e. for $n \in \mathbb{N}_0 = \mathbb{N} \cup \{0\}$ $\xi_n: \Omega \rightarrow I$ is \mathfrak{A} -measurable and for $i \in I$ P_i is a probability measure on (Ω, \mathfrak{A}) with $P_i(\xi_0 = i) = 1$ and $P_i(\xi_n = i_n | \xi_0 = i_0, \dots, \xi_{n-1} = i_{n-1}) = p_{i_{n-1}i_n}$ if the left side is defined. We assume that I is one recurrent class, i.e. for each $i \in I$ ξ_n visits each state infinitely often with P_i -probability 1. For a probability μ on I P_μ is the probability $\sum_{i \in I} \mu(i) P_i$ on (Ω, \mathfrak{A}) .

We fix once for all a distinguished point $O \in I$. Let $T_k: \Omega \rightarrow \mathbb{N}_0 \cup \{\infty\}$ be defined as follows:

$$T_0 = \inf \{n \geq 0: \xi_n = O\}$$
$$T_k = \inf \{n > T_{k-1}: \xi_n = O\}, \quad k \geq 1.$$

It is well known that for any starting probability μ and all $k \in \mathbb{N}_0$ $T_k < \infty$ P_μ -a.s., so the

$$\tau_k = T_k - T_{k-1}, \quad k \geq 1$$

are well defined if we restrict everything on a subspace of Ω which has full

measure for all P_μ . For the sake of notational convenience we write τ for τ_1 . The τ_k are well known to be independent and identically distributed.

If $f: I \rightarrow \mathbb{R}$, we call the sequence $f(\xi_0), f(\xi_1), \dots$ a functional of X .

If the chain is positive, i.e. $E_0(\tau) < \infty$, then there exists a unique stationary probability distribution $\Pi = (\pi(i))_{i \in I}$, i.e. we have $\sum_i \pi(i) p_{ij} = \pi(j)$ for all $j \in I$. ξ_0, ξ_1, \dots with the law P_π is then a stationary process. We call it the stationary chain.

In the sequel the chain X is assumed to be positive recurrent. The following central limit theorem is due to Doeblin (see [3], I. 16, Theorem 1).

Theorem A. *If the chain is positive and if $E_0 \left(\sum_{i=1}^{\tau} |f|(\xi_i) \right)^2 < \infty$ then $\Pi(f) = \sum_{i \in I} \pi(i) f(i)$ is well defined, and if*

$$\sigma^2(f) = E_0 \left(\sum_{n=1}^{\tau} (f(\xi_n) - \Pi(f)) \right)^2 > 0$$

then

$$\lim_{n \rightarrow \infty} P_0 \left(\frac{\sqrt{\alpha}}{\sigma \sqrt{n}} \sum_{j=1}^n (f(\xi_j) - \Pi(f)) \leq t \right) = \Phi(t)$$

where Φ is the standard normal distribution function and $\alpha = E_0(\tau)$.

Our main result is the following Berry-Esseen type bound:

Theorem 1. *Let μ be a starting probability on I . If*

$$E_0(\tau^3) < \infty \tag{1.1}$$

$$E_0 \left(\sum_{j=1}^{\tau} |f|(\xi_j) \right)^3 < \infty \tag{1.2}$$

$$E_\mu(T_0) < \infty \tag{1.3}$$

$$E_\mu \left(\sum_{j=1}^{T_0} |f|(\xi_j) \right) < \infty \tag{1.4}$$

then

$$\sup_t \left| P_\mu \left(\frac{\sqrt{\alpha}}{\sigma \sqrt{n}} \sum_{j=1}^n (f(\xi_j) - \Pi(f)) \leq t \right) - \Phi(t) \right| = O(n^{-1/2}) \tag{1.5}$$

The proof will be given in § 3.

Taking in particular $\mu = \Pi$, (1.3) and (1.4) are entailed by (1.1) and (1.2). To see this, the following result of Pitman [9] is useful:

Pitman's Occupation Measure Identity

Let $g: I^{\mathbb{N}_0} \rightarrow [0, \infty)$ be measurable, let S be a stopping time for ξ_0, ξ_1, \dots and let ν be the occupation measure on I defined by $\nu(i) = E_0 \left(\sum_{n=0}^{S-1} 1_i(\xi_n) \right)$. Then

$$E_0 \left(\sum_{n=0}^{S-1} g(\xi_n, \xi_{n+1}, \dots) \right) = \sum_{i \in I} v(i) E_i(g(\xi_0, \xi_1, \dots)).$$

With this result one easily proves the following

Lemma 1. *If $\mu = \Pi$ then (1.3) and (1.4) follow from (1.1) and (1.2).*

Proof. (1.3) is well known to follow from (1.1) (see e.g. [9]).

Let $S = \inf \{n > 0: \xi_n = 0\}$, $h(i) = \max(|f(i)|, 1)$. Then

$$\begin{aligned} E_\Pi \left(\sum_{j=1}^{T_0} |f|(\xi_j) \right) &\leq E_\Pi \left(\sum_{j=1}^S h(\xi_j) \right) \\ &\leq E_\Pi \left(\sum_{j=0}^{S-1} h(\xi_j) \right) + h(0) \\ &\leq E_\Pi \left(h(\xi_0) \sum_{j=0}^{S-1} h(\xi_j) \right) + h(0) \\ &= \pi(0) E_0 \left(\sum_{i=0}^{S-1} h(\xi_i) \sum_{j=1}^{S-1} h(\xi_j) \right) + h(0) \\ &\leq \pi(0) E_0 \left(\sum_{i=0}^{S-1} h(\xi_i) \right)^2 + h(0), \\ &\leq 2\pi(0) E_0 \left(\sum_{i=0}^{S-1} |f|(\xi_i) \right)^2 + 2\pi(0) E_0(\tau^2) + h(0), \end{aligned}$$

where the equality is by Pitman's identity, using the fact that the occupation measure for S is $\pi(0)\Pi$. So it is seen that (1.4) follows from (1.1) and (1.2).

From Lemma 1 and Theorem 1 one derives the following

Corollary 1. *If the starting probability μ is dominated by some multiple of Π and if (1.1) and (1.2) hold then (1.5) is true.*

It is desirable to have conditions based on more familiar entities. The following so-called strong mixing coefficients have been introduced by Rosenblatt (see [10]):

Let $\mathfrak{F}_k = \sigma(\xi_0, \dots, \xi_k)$ $\mathfrak{F}^k = \sigma(\xi_j, j \geq k)$. $\alpha(k)$, $k \geq 0$ is defined to be

$$\sup_{n \in \mathbb{N}_0} \sup_{A \in \mathfrak{F}_n} \sup_{B \in \mathfrak{F}^{n+k}} |P_\pi(A \cap B) - P_\pi(A)P_\pi(B)|$$

The following theorem will be proved in §4.

Theorem 2. *Let $\lambda \geq 0, \in \mathbb{R}$ then $\sum_{n=0}^{\infty} n^\lambda \alpha(n) < \infty$ if and only if the chain is aperiodic and $E_0(\tau^{\lambda+2}) < \infty$.*

With this result and Corollary 1 one has

Corollary 2. *If some multiple of Π dominates μ , if f is bounded and $\sum_n n \alpha(n) < \infty$ then (1.5) holds true.*

For unbounded functions one obtains for $p > 3$

$$\begin{aligned} E_0 \left(\sum_{j=1}^{\tau} |f|(\xi_j) \right)^3 &\leq E_0(\tau^{p-1} \sum_{j=1}^{\tau} |f|^p(\xi_j))^{3/p} \\ &\leq (E_0(\tau^{3(p-1)/(p-3)})^{(p-3)/p} \left(E_0 \left(\sum_{j=1}^{\tau} |f|^p \right) \right)^{3/p}. \end{aligned}$$

So one has

Corollary 3. *If some multiple of Π dominates μ and for a real number $p > 3$ $\Pi(|f|^p) < \infty$ and $\sum_n n^{(p+3)/(p-3)} \alpha(n) < \infty$ then (1.5) holds true.*

Bounds of order $O(n^{-1/2})$ for bounded functions f have been obtained by Lifshits [7] under conditions based on the maximum correlation coefficients, i.e. the cosine of the angle between the spaces $L_2(\mathfrak{F}_n)$ and $L_2(\mathfrak{F}^{n+k})$. Such conditions seem to be quite strong for Markov chains. If any of these angles is larger than zero $\alpha(n)$ converges to zero exponentially fast ([7], Theorem 5). It follows from our theorem 2 that for any chain with recurrence times with moments only of a finite order all maximal correlation coefficients equal 1.

The method of proof used here is the renewal approach which goes back to Doeblin:

Let $\rho_n = \max \{k: T_k \leq n\}$ and $l_n = T_{\rho_n}$; let further $X_n = \sum_{j=T_{n-1}+1}^{T_n} (f(\xi_j) - \Pi(f))$. The X_j are independent and identically distributed. Obviously

$$\begin{aligned} \sum_{j=1}^n (f(\xi_j) - \Pi(f)) &= \sum_{j=1}^{T_0} (f(\xi_j) - \Pi(f)) + \sum_{j=1}^{\rho_n} X_j \\ &\quad + \sum_{j=l_n+1}^n (f(\xi_j) - \Pi(f)). \end{aligned} \tag{1.6}$$

Theorem A then follows from the independence of the X_j , a central limit theorem with random summation and the asymptotic negligibility of first and third summand in (1.6) (after appropriate norming). However, error bounds of order $n^{-1/2}$ for central limit theorems with random summation are known only if X_j and ρ_n are independent, which certainly is not true in our case. Landers and Rogge in [5] derived bounds under rather general conditions, but applied to the situation in theorem 1 they only yield $O(n^{-1/4}(\log n)^{1/4})$ (see [6]). Bounds of order $O(n^{-1/3+\delta})$ under stronger conditions had previously been obtained by me with a modification of Landers' and Rogge's method [2]. Theorem 1 follows upon a close look at the dependence between ρ_n and the X_j .

A straightforward simplification of our proof also gives the following theorem which refutes the seemingly general belief that bounds of order $n^{-1/2}$ in central limit theorems with random summation are obtainable only in the independent case.

Theorem 3. *Let $(\eta_i, r_i)_{i \in \mathbb{N}}$ be independent identically distributed two-dimensional random variables with*

$$E(\eta_i) = 0, E(\eta_i^2) = 1, E(|\eta_i|^3) < \infty, \quad r_i \in \mathbb{N}, E(r_i^3) < \infty.$$

Let $\alpha = E(r_i)$ and $\rho_n = \max \left\{ k: \sum_{j=1}^k r_j \leq n \right\}$. Then

$$\sup_t \left| P\left(\sqrt{\alpha/n} \sum_{j=1}^{\rho_n} \eta_j \leq t\right) - \Phi(t) \right| = O(n^{-1/2}).$$

§ 2. A Semi-Local Berry-Esseen Bound

We prepare for the proof of Theorem 1 with a special Berry-Esseen theorem for two-dimensional i.i.d. random variables $((\zeta_n, \gamma_n), n \in \mathbb{N})$, which are lattice in one component. So we assume there is a $\rho \in \mathbb{R}$ such that $\gamma_n \in \rho + \mathbb{Z}$ a.s. It is further assumed that $E\zeta_n = E\gamma_n = 0, E|\zeta_n|^3 < \infty, E|\gamma_n|^3 < \infty$ and that the covariance matrix $\Sigma = (\sigma_{ij})_{i,j=1,2}$ has full rank 2.

Let $A = \{n \in \mathbb{N}: \exists k \in \mathbb{Z} \text{ with } P(\gamma - \rho = k) > 0, P(\gamma - \rho = k + n) > 0\}$. Clearly $A \neq \emptyset$, and for the sake of convenience we assume the largest common divisor d of A to be 1. This is not essential. The modifications needed in the case when this is not true are straightforward and therefore omitted.

Let φ be the two-dimensional density function of the centred normal distribution with covariance Σ , and let $\psi(x, y) = \int_{-\infty}^x \varphi(s, y) ds$. Let $S_n = \sum_{i=1}^n \zeta_i, T_n = \sum_{i=1}^n \gamma_i, \lambda_n(t_1, t_2)$ be the characteristic function of $(S_n/\sqrt{n}, T_n/\sqrt{n})$ and $g(t_1, t_2)$ be the characteristic function of $(\zeta_i, \gamma_i - \rho)$. Obviously

$$\lambda_n(t_1, t_2) = [g(t_1/\sqrt{n}, t_2/\sqrt{n}) \exp(it_2 \rho/\sqrt{n})]^n. \tag{2.1}$$

Lemma 1. *Given $\delta > 0$, there exist $\delta' > 0, 0 < r < 1$ and $C > 0$ such that $\lambda_n(t_1, t_2)$ and all partial derivatives up to the third (or any fixed) order are dominated in absolute value by Cr^{-n} for $|t_1| \leq \delta' \sqrt{n}, \delta \sqrt{n} \leq |t_2| \leq \pi \sqrt{n}$.*

Proof. From the assumption $d = 1$ it follows that $|g(0, v)|$ is bounded away from 1 uniformly in $\delta \leq |v| \leq \pi$. From continuity of g it follows that there is a $\delta' > 0, r < 1$ with $|g(u, v)| \leq r$ for $|u| \leq \delta', \delta \leq |v| \leq \pi$. The lemma now follows from (2.1) and the chain rule.

Proofs of the following two propositions may be found in [1] (Theorem 9.10 and Theorem 22.1).

Proposition A. *Let $\lambda_0(t_1, t_2) = \exp\left(-\frac{1}{2} \sum_{j,k=1}^2 t_j t_k \sigma_{jk}\right)$. There exist constants $\tilde{\epsilon}, \beta, c > 0$ (depending only on Σ and $E|\zeta_i|^3, E|\gamma_i|^3$) such that for $u, v \in \mathbb{N}_0, u + v \leq 3$*

$$\left| \frac{\partial^{u+v}}{\partial t_1^u \partial t_2^v} (\lambda_n(t_1, t_2) - \lambda_0(t_1, t_2)) \right| \leq \frac{c}{\sqrt{n}} |t|^{3-u-v} e^{-\beta|t|^2}$$

for $|t_1|, |t_2| \leq \tilde{\varepsilon} \sqrt{n}$, where $t = (t_1, t_2)$ and $|t| = (t_1^2 + t_2^2)^{1/2}$.

For $\alpha \in \mathbb{Z}$ let $y_{\alpha, n} = (n\rho + \alpha)/\sqrt{n}$.

Proposition B.

$$\sup_{\alpha \in \mathbb{Z}} (1 + |y_{\alpha, n}|^3) |P(T_n/\sqrt{n} = y_{\alpha, n}) - \psi(\infty, y_{\alpha, n})/\sqrt{n}| = O(n^{-1}).$$

The main result of this section is

Theorem 4. *Under the above stated conditions*

$$\sup_{x \in \mathbb{R}} \sup_{\alpha \in \mathbb{Z}} (1 + y_{\alpha, n}^2) \left| P \left(\frac{S_n}{\sqrt{n}} \leq x, \frac{T_n}{\sqrt{n}} = y_{\alpha, n} \right) - \frac{1}{\sqrt{n}} \psi(x, y_{\alpha, n}) \right| = O \left(\frac{1}{n} \right).$$

Remark. The proof given below easily gives the stronger statement where $y_{\alpha, n}^2$ is replaced by $|y_{\alpha, n}|^{3-\delta}$ ($\delta > 0$). The statement with $|y_{\alpha, n}|^3$ may also be true but would probably require more refined techniques. The above theorem is sufficient for our purpose.

Proof. Let $F_n(x, y_{\alpha, n}) = P(S_n/\sqrt{n} \leq x, T_n/\sqrt{n} = y_{\alpha, n})$ and $\hat{F}_n(x, y_{\alpha, n}) = F_n(x, y_{\alpha, n})/F_n(\infty, y_{\alpha, n})$ if $F_n(\infty, y_{\alpha, n}) > 0$, $\hat{F}_n(x, y_{\alpha, n}) = 1_{[0, \infty)}(x)$ if $F_n(\infty, y_{\alpha, n}) = 0$. For fixed $\alpha \in \mathbb{Z}$ $\hat{F}_n(\cdot, y_{\alpha, n})$ is a distribution function. Let $\hat{\psi}(x, y_{\alpha, n}) = \psi(x, y_{\alpha, n})/\psi(\infty, y_{\alpha, n})$. For $T > 0$ let $v_T(x) = (1 - \cos(Tx))/(\pi T x^2)$. v_T is the density of a probability distribution with characteristic function $\omega_T(\lambda) = \max(0, 1 - |\lambda|/T)$.

Let $F_n^T(x, y) = \int_{-\infty}^{\infty} F_n(x-u, y) v_T(u) du$ and $\hat{F}_n^T, \psi^T, \hat{\psi}^T$ be defined by similar convolutions. (We drop indices α, n in $y_{\alpha, n}$ for the sake of notational simplicity.)

From Lemma 3.1, Ch. XVI of [4]

$$\begin{aligned} & \sup_x |\hat{F}_n(x, y) - \hat{\psi}(x, y)| \\ & \leq 2 \sup_x |\hat{F}_n^T(x, y) - \hat{\psi}^T(x, y)| + \frac{12}{\pi T} \sup_x \left| \frac{\partial}{\partial x} \hat{\psi}(x, y) \right|. \end{aligned}$$

After some elementary calculations it follows that

$$\begin{aligned} \sup_x |F_n(x, y) - \psi(x, y)/\sqrt{n}| & \leq 2 \sup_x |F_n^T(x, y) - \psi^T(x, y)/\sqrt{n}| \\ & + 3 |F_n(\infty, y) - \psi(\infty, y)/\sqrt{n}| + 12 \sup_x \varphi(x, y)/(\pi T \sqrt{n}). \end{aligned}$$

Combining this with Proposition B and taking $T \sim \sqrt{n}$ one has

$$\begin{aligned} & \sup_{x \in \mathbb{R}, \alpha \in \mathbb{Z}} (1 + y_{\alpha, n}^2) |F_n(x, y_{\alpha, n}) - \psi(x, y_{\alpha, n})/\sqrt{n}| \\ & \leq 2 \sup_{x, \alpha} (1 + y_{\alpha, n}^2) |F_n^T(x, y_{\alpha, n}) - \psi^T(x, y_{\alpha, n})/\sqrt{n}| + O \left(\frac{1}{n} \right). \end{aligned} \tag{2.2}$$

From now on we take $T = \varepsilon \sqrt{n}$, $\varepsilon = \min(\tilde{\varepsilon}, \delta')$ where $\tilde{\varepsilon}$ comes from Proposition A, δ' from lemma 1, and in this lemma $\delta = \tilde{\varepsilon}$. Now

$$F_n^T(x, y) - \psi^T(x, y)/\sqrt{n} = \frac{1}{(2\pi)^2 \sqrt{n}} \int_{-\varepsilon\sqrt{n}}^{\varepsilon\sqrt{n}} \int_{-\pi\sqrt{n}}^{\pi\sqrt{n}} \frac{1}{-it_1} e^{-it_1 x - it_2 y} \omega_T(t_1) \cdot (\lambda_n(t_1, t_2) - \lambda_0(t_1, t_2)) dt_2 dt_1 \quad (2.3)$$

and therefore if $z < x$.

$$\begin{aligned} & y^2(F_n^T(x, y) - \psi^T(x, y)/\sqrt{n} - (F_n^T(z, y) - \psi^T(z, y)/\sqrt{n})) \\ &= \frac{1}{(2\pi)^2 \sqrt{n}} \int_{-\varepsilon\sqrt{n}}^{\varepsilon\sqrt{n}} \int_{-\pi\sqrt{n}}^{\pi\sqrt{n}} \frac{1}{it_1} (e^{-it_1 x} - e^{-it_1 z}) e^{-it_2 y} \\ & \quad \cdot \omega_T(t_1) \frac{\partial^2}{\partial t_2^2} (\lambda_n(t_1, t_2) - \lambda_0(t_1, t_2)) dt_2 dt_1 \\ &= \frac{1}{(2\pi)^2 \sqrt{n}} \int_{-\varepsilon\sqrt{n}}^{\varepsilon\sqrt{n}} \left\{ \int_{-\tilde{\varepsilon}\sqrt{n}}^{\tilde{\varepsilon}\sqrt{n}} \dots \frac{\partial^2}{\partial t_2^2} (\lambda_n(t_1, t_2) - \lambda_0(t_1, t_2)) dt_2 \right. \\ & \quad + \int_{|t_2| \in \sqrt{n}[\tilde{\varepsilon}, \pi]} \dots \frac{\partial^2}{\partial t_2^2} \lambda_n(t_1, t_2) dt_2 \\ & \quad \left. - \int_{|t_2| \in \sqrt{n}[\tilde{\varepsilon}, \pi]} \dots \frac{\partial^2}{\partial t_2^2} \lambda_0(t_1, t_2) dt_2 \right\} dt_1 \\ &= I_1 + I_2 + I_3 \quad \text{say.} \end{aligned} \quad (2.4)$$

We write

$$\begin{aligned} h(t_1, t_2) &= \frac{\partial^2}{\partial t_2^2} (\lambda_n(t_1, t_2) - \lambda_0(t_1, t_2)) \\ h(t_1, t_2) &= (1 - e^{-\beta t_1^2}) h(t_1, t_2) + e^{-\beta t_1^2} (h(t_1, t_2) - h(0, t_2)) \\ & \quad + e^{-\beta t_1^2} h(0, t_2), \end{aligned} \quad (2.5)$$

where β is from Proposition A. From this proposition one has for $|t_1| \leq \varepsilon$, $|t_2| \leq \tilde{\varepsilon}$:

$$\begin{aligned} |h(t_1, t_2)| &\leq \frac{c}{\sqrt{n}} |t| e^{-\beta |t|^2}, \quad |h(t_1, t_2) - h(0, t_2)| \leq \frac{c}{\sqrt{n}} |t_1| e^{-\beta t_1^2}, \\ |h(0, t_2)| &\leq \frac{c}{\sqrt{n}} e^{-\beta t_2^2}. \end{aligned}$$

Further $|1 - e^{-\beta t_1^2}| \leq \beta t_1^2$. Implementing these estimates in (2.4) and (2.5) one has

$$\begin{aligned} |I_1| &\leq \frac{c'}{\sqrt{n}} \left\{ \int_{-\varepsilon\sqrt{n}}^{\varepsilon\sqrt{n}} \left\{ \int_{-\tilde{\varepsilon}\sqrt{n}}^{\tilde{\varepsilon}\sqrt{n}} |t_1| \omega_T(t_1) e^{-\beta |t|^2} dt_2 \right. \right. \\ & \quad \left. \left. + \int_{-\tilde{\varepsilon}\sqrt{n}}^{\tilde{\varepsilon}\sqrt{n}} |\omega_T(t_1)| e^{-\beta |t|^2} dt_2 \right\} dt_1 \right. \\ & \quad \left. + \left| \int_{-\varepsilon\sqrt{n}}^{\varepsilon\sqrt{n}} \frac{1}{it_1} (e^{-it_1 x} - e^{-it_1 z}) \omega_T(t_1) e^{-\beta t_1^2} dt_1 \right| \int_{-\tilde{\varepsilon}\sqrt{n}}^{\tilde{\varepsilon}\sqrt{n}} e^{-\beta t_2^2} dt_2 \right\} \\ &= O(n^{-1}) \end{aligned}$$

uniformly in x, z, α ,

I_2 can be handled similarly by using Lemma 1 instead of Proposition B and splitting as follows:

$$\frac{\partial^2}{\partial t_2^2} \lambda_n(t_1, t_2) = \frac{\partial^2}{\partial t_2^2} (\lambda_n(t_1, t_2) - \lambda_n(0, t_2)) + \frac{\partial^2}{\partial t_2^2} \lambda_n(0, t_2).$$

In this way one obtains $|I_2| = O(\delta^{-n})$ and in the same way, using exponential decrease of λ_0 , one has exponential decrease of $|I_3|$. So $|I_1| + |I_2| + |I_3| = O(n^{-1})$, and letting $z \rightarrow -\infty$ one has

$$y^2(F_n^T(x, y) - \psi^T(x, y)/\sqrt{n}) = O(n^{-1}).$$

In the same way

$$(F_n^T(x, y) - \psi^T(x, y)/\sqrt{n}) = O(n^{-1})$$

and from these estimates and (2.2) the theorem follows.

§ 3. Proof of Theorem 1

In this section $c, c', \varepsilon, \varepsilon'$ are always constants > 0 , c, c' “sufficiently large” and $\varepsilon, \varepsilon'$ “sufficiently small” which do not depend on n, m, t, s etc. They may vary from formula to formula but not in the same.

We resume the notation of § 1. We have from (1.6)

$$\left\{ \frac{\sqrt{\alpha}}{\sigma\sqrt{n}} \sum_{j=1}^n (f(\xi_j) - \Pi(f)) \leq t \right\} \\ = \bigcup_{m=0}^n \bigcup_{s=0}^n \bigcup_{r=0}^n \left\{ \frac{\sqrt{\alpha}}{\sigma\sqrt{n}} \left(\sum_{j=1}^r (f(\xi_j) - \Pi(f)) + \sum_{j=1}^m X_j \right. \right. \\ \left. \left. + \sum_{j=n-s+1}^m (f(\xi_j) - \Pi(f)) \right) \leq t, T_0 = r, \sum_{j=1}^m \tau_j = n - s - r, \tau_{m+1} > s \right\}.$$

By the Markov property

$$P_\mu \left(\frac{\sqrt{\alpha}}{\sigma\sqrt{n}} \sum_{j=1}^n (f(\xi_j) - \Pi(f)) \leq t \right) \\ = \sum_{m=0}^n \sum_{s=0}^n \sum_{r=0}^n \iint P_0 \left(\frac{\sqrt{\alpha}}{\sigma\sqrt{n}} \sum_{j=1}^m X_j \leq t - u - v, \sum_{j=1}^m \tau_j = n - s - r \right) \\ \cdot P_\mu(R_r \in dv, T_0 = r) P_0(R_s \in du, \tau > s) \tag{3.1}$$

where

$$R_k = \frac{\sqrt{\alpha}}{\sigma\sqrt{n}} \sum_{j=1}^k (f(\xi_j) - \Pi(f)).$$

The sum on the right side of (3.1) may be splitted as

$$\sum_{s=0}^{\sqrt{n}} \sum_{r=0}^{\sqrt{n}} \sum_{m=0}^n + \sum_{s=\sqrt{n}+1}^n \sum_{r=0}^n \sum_{m=0}^n + \sum_{r=\sqrt{n}+1}^n \sum_{s=0}^n \sum_{m=0}^n$$

where it is understood that summation begins or ends at the integer part of a number. The second summand is bounded by

$$\sum_{s=\sqrt{n}+1}^{\infty} P_0(\tau > s) = O(n^{-1})$$

and the third by

$$\sum_{r=\sqrt{n}+1}^{\infty} P_{\mu}(T_0 = r) = O(n^{-1/2})$$

so in order to prove the theorem it suffices to consider summation over s, r up to \sqrt{n} in (3.1). Clearly $m=0$ may then be excluded.

Let $\zeta_k = X_k/\sigma$, $\gamma_k = \tau_k - \alpha$. For the moment we assume that (ζ_k, τ_k) has covariance matrix of rank 2, so we can apply theorem 4 of §2. We assumed there $d = 1$ which means that the chain is aperiodic. However, this is only for notational convenience and is easily seen to be of no importance. We have

$$\begin{aligned} P_{\mu} \left(\frac{\sqrt{\alpha}}{\sigma\sqrt{n}} \sum_{j=1}^m X_j \leq t - u - v, \sum_{j=1}^m \tau_j = n - s - r \right) \\ = \frac{1}{\sqrt{m}} \psi \left(\sqrt{\frac{n}{\alpha m}} (t - u - v), \lambda_{r,s,m} \right) + O \left(\frac{1}{m(1 + \lambda_{r,s,m}^2)} \right) \end{aligned} \tag{3.3}$$

for $m \geq 1$ where $\lambda_{r,s,m} = (n - s - r - \alpha m) / \sqrt{m}$.

The O -term on the right side of (3.3) does not depend on u, v , so from (3.1)–(3.3) follows

$$\begin{aligned} P_{\mu} \left(\frac{\sqrt{\alpha}}{\sigma\sqrt{n}} \sum_{j=1}^n (f(\xi_j) - \Pi(f)) \leq t \right) \\ = \sum_{s=0}^{\sqrt{n}} \sum_{r=0}^{\sqrt{n}} \sum_{m=1}^n \left\{ \iint \psi \left(\sqrt{\frac{n}{\alpha m}} (t - u - v), \lambda_{r,s,m} \right) \right. \\ \cdot P_{\mu}(R_r \in dv, T_0 = r) P_0(R_s \in du, \tau_1 > s) \\ \left. + O \left(\frac{1}{m} (1 + \lambda_{r,s,m}^2)^{-1} \right) P_{\mu}(T_0 = r) P_0(\tau_1 > s) \right\}. \end{aligned} \tag{3.4}$$

The theorem then follows from the following three relations

$$\sum_{s=0}^{\sqrt{n}} \sum_{r=0}^{\sqrt{n}} \sum_{m=1}^n O \left(\frac{1}{m} (1 + \lambda_{r,s,m}^2)^{-1} \right) = O(n^{-1/2}) \tag{3.5}$$

$$\begin{aligned} & \sum_{s=0}^{\sqrt{n}} \sum_{r=0}^{\sqrt{n}} \sum_{m=1}^n \left| \iint \frac{1}{\sqrt{m}} \psi \left(\sqrt{\frac{n}{\alpha m}} (t-u-v), \lambda_{r,s,m} \right) \right. \\ & \quad \cdot P_0(R_s \in du, \tau > s) P_\mu(R_r \in dv, T_0 = r) \\ & \quad \left. - \frac{1}{\sqrt{m}} \psi(t, \lambda_{r,s,m}) P_0(\tau > s) P_\mu(T_0 = r) \right| = O(n^{-1/2}) \end{aligned} \quad (3.6)$$

$$\sum_{s=0}^{\sqrt{n}} \sum_{r=0}^{\sqrt{n}} \left| \sum_{m=1}^n \frac{1}{\sqrt{m}} \psi(t, \lambda_{r,s,m}) - \frac{1}{\alpha} \int_{-\infty}^{\infty} \psi(t, x) dx \right| P_0(\tau > s) P_\mu(T_0 = r) = O(n^{-1/2}) \quad (3.7)$$

everything uniformly in t .

Indeed (3.3)-(3.7) imply

$$\begin{aligned} & P \left(\frac{\sqrt{\alpha}}{\sigma \sqrt{n}} \sum_{j=1}^n (f(\xi_j) - \Pi(f)) \leq t \right) \\ & = \left(\sum_{s=0}^{\sqrt{n}} \sum_{r=0}^{\sqrt{n}} \frac{1}{\alpha} P_0(\tau > s) P_\mu(T_0 = r) \right) \int_{-\infty}^{\infty} \psi(t, x) dx + O(n^{-1/2}) \\ & = \phi(t) + O(n^{-1/2}). \end{aligned}$$

So the theorem is proved in the case where the covariance matrix of (ζ, γ) is nondegenerated. It remains to prove (3.5)-(3.7) for this case.

Proof of (3.5). For $r, s \leq \sqrt{n}$

$$\frac{1}{m} (1 + \lambda_{r,s,m}^2)^{-1} \leq \begin{cases} 1/m \leq c/n & \text{for } |n - \alpha m| \leq 2\sqrt{n} \\ (n - \alpha m)^{-2} & \text{for } \alpha m > n + 2\sqrt{n} \\ (n - 2\sqrt{n} - \alpha m)^{-2} & \text{for } \alpha m < n - 2\sqrt{n} \end{cases}$$

From this (3.5) follows by some elementary calculations

Proof of (3.6). Let $I_m(t, u, v)$ be the interval between t and $\sqrt{\frac{n}{\alpha m}}(t-u-v)$. We have

$$\begin{aligned} & \left| \psi \left(\sqrt{\frac{n}{\alpha m}} (t-u-v), \lambda_{r,s,m} \right) - \psi(t, \lambda_{r,s,m}) \right| \\ & \leq \left(\sqrt{\frac{n}{\alpha m}} (|u| + |v|) \sup_{x \in \mathbb{R}} \varphi(x, \lambda_{r,s,m}) \right) \\ & \quad + |t| \left| \sqrt{\frac{n}{\alpha m}} - 1 \right| \sup_{x \in I_m} \varphi(x, \lambda_{r,s,m}). \end{aligned}$$

So

$$\begin{aligned} & \sum_{s=0}^{\sqrt{n}} \sum_{r=0}^{\sqrt{n}} \sum_{m=1}^n \left| \iint \frac{1}{\sqrt{m}} \psi \left(\sqrt{\frac{n}{\alpha m}} (t-u-v), \lambda_{r,s,m} \right) \right. \\ & \quad \cdot P_0(R_s \in du, \tau > s) P_\mu(R_r \in dv, T_0 = r) \\ & \quad \left. - \frac{1}{\sqrt{m}} \psi(t, \lambda_{r,s,m}) P_0(\tau > s) P_\mu(T_0 = r) \right| \end{aligned}$$

$$\begin{aligned}
 &\leq \sum_{s=0}^{\sqrt{n}} \sum_{r=0}^{\sqrt{n}} \sum_{m=1}^n \left\{ \int \frac{1}{\sqrt{m}} \sqrt{\frac{n}{\alpha m}} |u| \sup_{x \in \mathbb{R}} \varphi(x, \lambda_{r,s,m}) P_0(R_s \in du, \tau > s) P_\mu(T_0 = r) \right. \\
 &\quad + \int \frac{1}{\sqrt{m}} \sqrt{\frac{n}{\alpha m}} |v| \sup_{x \in \mathbb{R}} \varphi(x, \lambda_{r,s,m}) P_0(\tau > s) P_\mu(R_r \in dv, T = r) \\
 &\quad + \iint \frac{1}{\sqrt{m}} |t| \left| \sqrt{\frac{n}{\alpha m}} - 1 \right| \sup_{x \in I_m(t, u, v)} |\varphi(x, \lambda_{r,s,m})| P_0(R_s \in du, \tau > s) \\
 &\quad \cdot P_\mu(R_r \in dv, T_0 = r) \\
 &= A_1 + A_2 + A_3 \quad \text{say.}
 \end{aligned} \tag{3.8}$$

Obviously $\sup_{x \in \mathbb{R}} \varphi(x, \lambda) \leq c \exp(-\varepsilon \lambda^2)$ and on $\{\tau > s\}$

$$|R_s| \leq \frac{\sqrt{\alpha}}{\sigma \sqrt{n}} \sum_{j=1}^{\tau} |f(\xi_j) - \Pi(f)| = Z \quad \text{say.}$$

So

$$\begin{aligned}
 \int |u| P_0(R_s \in du, \tau > s) &= E_0(|R_s| 1_{\{\tau > s\}}) \leq E(Z 1_{\{\tau > s\}}) \\
 &\leq \frac{1}{s^2} E(Z \tau^2) \leq c/(\sqrt{n} s^2)
 \end{aligned}$$

by Hölder's inequality. So

$$A_1 \leq c \sum_{s=0}^{\sqrt{n}} \sum_{r=0}^{\sqrt{n}} \left(\frac{1}{s^2} \wedge 1 \right) P_\mu(T_0 = r) \sum_{m=1}^n \frac{1}{m} \exp(-\varepsilon \lambda_{r,s,m}^2). \tag{3.9}$$

Let now

$$\begin{aligned}
 A_0 &= \{m: n - 3\sqrt{n} \leq \alpha m \leq n + \sqrt{n}\} \\
 A_k &= \{m: n + k\sqrt{n} < \alpha m \leq n + (k+1)\sqrt{n}\}; \quad k \geq 1 \\
 A'_k &= \{m: n - (k+1)\sqrt{n} \leq \alpha m < n - k\sqrt{n}\}; \quad k \geq 3.
 \end{aligned}$$

Remarking now that for $s, r \leq \sqrt{n}$, $m \leq n$, $m \in A_k$ one has $|\lambda_{r,s,m}| \geq \frac{k}{\sqrt{2}}$, and for $m \in A'_k$ $|\lambda_{r,s,m}| \geq (k-2)$ one obtains by splitting the sum $\sum_{m=1}^n$ into the sums over the A 's after some elementary calculations

$$\sum_{m=1}^n \frac{1}{m} \exp(-\varepsilon \lambda_{r,s,m}^2) = O(n^{-1/2}) \quad \text{for } s, r \leq \sqrt{n},$$

so $A_1 = O(n^{-1/2})$ follows. A_2 can be handled similarly. We consider now A_3 .

Let Z be as above and

$$Z' = \frac{\sqrt{\alpha}}{\sigma \sqrt{n}} \sum_{j=1}^{T_0} |f(\xi_j) - \Pi(f)|.$$

For $|u| + |v| \leq \frac{2|t|}{3}$

$$\sup_{x \in I_m(t, u, v)} \varphi(x, \lambda_{r,s,m}) \leq c \exp\left(-\varepsilon \frac{n}{\alpha m} t^2\right) \exp(-\varepsilon \lambda_{r,s,m}^2).$$

Therefrom

$$\begin{aligned} & \sup_{t \in \mathbb{R}} \iint |t| \sup_{x \in I_m(t, u, v)} \varphi(x, \lambda_{r, s, m}) P_0(R_s \in du, Z \leq \frac{|t|}{3}, \tau > s) \\ & \cdot P_\mu \left(R_r \in dv, Z' \leq \frac{|t|}{3}, T_0 = r \right) \end{aligned} \quad (3.10)$$

$$\begin{aligned} & \leq c \sqrt{\frac{\alpha m}{n}} \exp(-\varepsilon \lambda_{r, s, m}^2) P_0(\tau > s) P_\mu(T_0 = r) \\ & \sup_{t \in \mathbb{R}} |t| P_\mu \left(Z' > \frac{|t|}{3}, T_0 = r \right) \leq c E_\mu(Z' 1_{T_0=r}) \end{aligned} \quad (3.11)$$

$$\sup_{t \in \mathbb{R}} |t| P_0 \left(Z > \frac{|t|}{3}, \tau > s \right) \leq c \left(\frac{1}{s^2} \wedge 1 \right). \quad (3.12)$$

Combining (3.10)–(3.12) gives

$$\begin{aligned} A_3 & \leq c \sum_{s=0}^{\sqrt{n}} \sum_{r=0}^{\sqrt{n}} \sum_{m=1}^n \frac{1}{\sqrt{m}} \left| \sqrt{\frac{n}{\alpha m}} - 1 \right| \exp(-\varepsilon \lambda_{r, s, m}^2) \\ & \cdot \left(\left(\frac{1}{s^2} \wedge 1 \right) (E_\mu(Z' 1_{T_0=r}) + P_\mu(T_0=r)) \right). \end{aligned}$$

Splitting the sum over m into subsummations over the A 's one obtains after some elementary calculations for $s, r \leq \sqrt{n}$

$$\sum_{m=1}^n \frac{1}{\sqrt{m}} \left| \sqrt{\frac{n}{\alpha m}} - 1 \right| \exp(-\varepsilon \lambda_{r, s, m}^2) = O(n^{-1/2}).$$

So $A_3 = O(n^{-1/2})$ follows.

Proof of (3.7). For fixed s, r $\lambda_{r, s, m}$ decreases as m increases and

$$\begin{aligned} \lambda_{r, s, m} - \lambda_{r, s, m+1} & = \alpha \frac{1}{\sqrt{m}} + \lambda_{r, s, m+1} \left(\underbrace{\sqrt{1 + \frac{1}{m}} - 1}_{= O\left(\frac{1}{m}\right)} \right) \\ & = O\left(\frac{1}{m}\right). \end{aligned}$$

From this one easily derives

$$\begin{aligned} & \sum_{s=0}^{\sqrt{n}} \sum_{r=0}^{\sqrt{n}} \left| \frac{1}{\sqrt{m}} \psi(t, \lambda_{r, s, m}) - \frac{1}{\alpha} (\lambda_{r, s, m} - \lambda_{r, s, m+1}) \psi(t, \lambda_{r, s, m}) \right| \\ & \cdot P_0(\tau > s) P_\mu(T_0 = r) = O(n^{-1/2}). \end{aligned} \quad (3.13)$$

Further

$$\sup_{x \in [\lambda_{m+1}, \lambda_m]} |\psi(t, x) - \psi(t, \lambda_m)| \leq c(\lambda_m - \lambda_{m+1}) \exp(-\varepsilon \lambda_m^2)$$

and therefore

$$\begin{aligned} & \sum_{s=0}^{\sqrt{n}} \sum_{r=0}^{\sqrt{n}} \left| \sum_{m=1}^n \frac{1}{\alpha} (\lambda_{r,s,m} - \lambda_{r,s,m+1}) \psi(t, \lambda_{r,s,m}) \right. \\ & \quad \cdot P_0(\tau > s) P_\mu(T_0 = r) - \int_{\lambda_{r,s,n}}^{\lambda_{r,s,1}} \psi(t, x) dx P_0(\tau > s) P_\mu(T_0 = r) \left. \right| \\ & \leq c \sum_{s=0}^{\sqrt{n}} \sum_{r=0}^{\sqrt{n}} \sum_{m=1}^n (\lambda_{r,s,m} - \lambda_{r,s,m+1})^2 \exp(-\varepsilon \lambda_{r,s,m}^2) \\ & \quad \cdot P_0(\tau > s) P_\mu(T_0 = r) = O(n^{-1/2}). \end{aligned}$$

Obviously

$$\int_{\lambda_n}^{\lambda_1} \psi(t, x) dx = \int_{-\infty}^{\infty} \psi(t, x) dx + O(n^{-1/2}),$$

so (3.13)–(3.14) entail (3.7).

The case where Σ is degenerated is much more simple. First, if τ is norandom it can easily be reduced to the standard Berry-Esseen theorem. If τ is nondegenerated but Σ has rank 1, then there exists a constant $a \in \mathbb{R}$ such that $\zeta_i = a\gamma_i$ a.s. A typical example for this is if $f = 1_{\{0\}}$. In this special case the statement of theorem 1 (with fixed starting point) has been proved by Landers and Rogge in [6] (theorem 1). Their proof can easily be adapted to the general case where $\zeta_i = a\gamma_i$. We omit the details.

§ 4. Proof of Theorem 2

Rosenblatt ([10], VII.3, Lemma 1) obtained the result that a Markov chain is strongly mixing, i.e. $\lim_{n \rightarrow \infty} \alpha(n) = 0$, if and only if

$$\sup_{i \in I} \left\{ \sum \pi(i) |E_i(f(\xi_n)) - \Pi(f)| : f: I \rightarrow \mathbb{R}, \|f\|_\infty \leq 1 \right\}$$

goes to 0 as $n \rightarrow \infty$. His proof easily gives the following stronger statement:

Lemma 2. $\alpha(n) \leq \frac{1}{2} \sup_{\|f\|_\infty \leq 1} \Pi(|E \cdot (f(\xi_n)) - \Pi(f)|)$
 $\leq 2 \sup_{A, B \subset I} |P_\Pi(\xi_0 \in A, \xi_n \in B) - \pi(A)\pi(B)|$
 $\leq 2\alpha(n).$

Proof of Theorem 2. (I) It is assumed that $\sum_{n=1}^{\infty} n^p \alpha(n) < \infty$ for some $p \geq 0$. Let $A_n = \{\xi_j \neq 0 \text{ for } j = n+1, n+2, \dots, 2n\}$. For any starting probability μ

$$P_\mu(A_n) = \sum_{m=1}^n P_\mu(\xi_m = 0) P_0(S > 2n - m) + P_\mu(S > 2n).$$

Taking $\mu = \delta_0$ and $\mu = \Pi$ one obtains

$$|P_0(S > 2n) - P_\pi(S > 2n)| \\ \leq |P_0(A_n) - P_\Pi(A_n)| + P_0(S > n) \sum_{m=1}^{\infty} |P_0(\xi_m = 0) - \pi(0)|.$$

Now $|P_0(A_n) - P_\Pi(A_n)| \leq \alpha(n)/\pi(0)$ and $|P_0(\xi_m = 0) - \pi(0)| \leq \alpha(m)/\pi(0)$. So for q , $0 \leq q \leq p$,

$$\pi(0) \sum_{n=1}^{\infty} n^q |P_0(S > 2n) - P_\Pi(S > 2n)| \\ \leq \sum_{n=1}^{\infty} n^q \alpha(n) + \left(\sum_{n=1}^{\infty} \alpha(n) \right) \left(\sum_{n=1}^{\infty} n^q P_0(S > n) \right).$$

So it follows that if $E_0(S^{q+1}) < \infty$ then $E_\Pi(S^{q+1}) < \infty$. On the other hand it is well known that for any $r \geq 0$ $E_0(S^{r+1}) < \infty$ if and only if $E_\Pi(S^r) < \infty$. So it clearly follows that $E_0(S^{p+2}) < \infty$.

(II) Let us prove the converse, so we assume $E_0(S^{p+2}) < \infty$ for some $p \geq 0$ or, what is the same, $E_\Pi(S^{p+1}) < \infty$.

We use the Pitman coupling technique (see [8]), so let ξ_n, ξ'_n be two independent chains with transition probabilities p_{ij} . We write \hat{P} , for the law of the pair (ξ_n, ξ'_n) . Let $R = \inf \{n \geq 0: \xi_n = 0, \xi'_n = 0\}$ and let $f: I \rightarrow \mathbb{R}$, $\|f\|_\infty \leq 1$. As in Pitman [8]

$$|E_i(f(\xi_n)) - E_\Pi(f(\xi_n))| \leq 2\hat{P}_{\delta_i \times \Pi}(R \geq n)$$

so

$$\Pi(|E_\cdot(f(\xi_n)) - \Pi(f)|) \leq 2\hat{P}_{\Pi \times \Pi}(R \geq n).$$

If $E_\Pi(S^{p+1}) < \infty$ Pitman proved in [8] that $E_{\Pi \times \Pi}(R^{p+1}) < \infty$ so $\sum_{n=1}^{\infty} n^p \hat{P}_{\Pi \times \Pi}(R \geq n) < \infty$ and $\sum_{n=1}^{\infty} n^p \alpha(n) < \infty$ follows from Lemma 2.

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