

**LARGE-DEVIATION PROBABILITIES FOR
ONE-DIMENSIONAL MARKOV CHAINS.
PART 3: PRESTATIONARY DISTRIBUTIONS IN THE
SUBEXPONENTIAL CASE***

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(Translated by D. A. Korshunov)

Abstract. This paper continues investigations of A. A. Borovkov and D. A. Korshunov [*Theory Probab. Appl.*, 41 (1996), pp. 1–24 and 45 (2000), pp. 379–405]. We consider a time-homogeneous Markov chain $\{X(n)\}$ that takes values on the real line and has increments which do not possess exponential moments. The asymptotic behavior of the probability $\mathbf{P}\{X(n) \geq x\}$ is studied as $x \rightarrow \infty$ for fixed values of time n and for unboundedly growing n as well.

Key words. Markov chain, asymptotic behavior of large-deviation probabilities, subexponential distribution, invariant measure, integrated distribution tail

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1. Introduction. Let $X(n) = X(y, n)$, $n = 0, 1, \dots$, be a time-homogeneous Markov chain with values on the real line \mathbf{R} and with initial state $y \equiv X(y, 0)$. Denote by $P(y, B) = \mathbf{P}\{X(y, 1) \in B\}$, where B is a Borel set in \mathbf{R} , the transition probability of the chain.

Let $\xi(y)$ be the increment of the chain X in one step at point $y \in \mathbf{R}$, that is, $\xi(y) = X(y, 1) - y$.

One of the main objects under study in the present paper consists of *asymptotically homogeneous* in space Markov chains, that is, chains for which the distribution of $\xi(y)$ converges weakly as $y \rightarrow \infty$ to the distribution of some random variable ξ ; we denote it by $\xi(y) \Rightarrow \xi$. We assume everywhere that $m = \mathbf{E}\xi < 0$ and $\mathbf{P}\{\xi > 0\} > 0$. Also, we extend our analysis to a more general class of chains with asymptotically homogeneous drift, that is, chains which are satisfying only the property $\mathbf{E}\xi(y) \rightarrow m$ as $y \rightarrow \infty$.

In [2], we classified the asymptotically homogeneous-in-space Markov chains with respect to the behavior of the Laplace transform $\varphi(\lambda) = \mathbf{E}e^{\lambda\xi}$. In terms of that classification, in the present paper we investigate the third case (c) which corresponds to the situation $\varphi(\lambda) = \infty$ for any $\lambda > 0$. Of course, we assume regular behavior of the tail $\mathbf{P}\{\xi \geq x\}$ of the distribution ξ as $x \rightarrow \infty$. Let us recall some definitions.

A positive function g is called *long-tailed* if, for any fixed t , the limit of the ratio $g(u+t)/g(u)$ is equal to 1 as $u \rightarrow \infty$. We say that a *distribution G is long-tailed* if the tail $G(x) \equiv G([x, \infty))$ of this distribution is long-tailed.

Note that, for any random variable ξ with long-tailed distribution, $\mathbf{E}e^{\lambda\xi} = \infty$ for any $\lambda > 0$.

We say [3] that a distribution G on \mathbf{R}^+ with unbounded support belongs to the class \mathcal{S} (and is called a *subexponential* distribution) if the convolution $G * G$ satisfies

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the equivalence $G * G(x) \sim 2G(x)$ as $x \rightarrow \infty$.

It is shown in [3] that any subexponential distribution G is long-tailed with necessity. Sufficient conditions for subexponentiality may be found, for example, in [3], [8]. In particular, the class \mathcal{S} includes distributions with the tail $G(x) = x^{-\alpha} \varepsilon(x)$, where $\alpha > 0$ and $\varepsilon(x)$ is a slowly-varying-at-infinity function. Moreover, the class \mathcal{S} also contains the so-called *upper power* distributions, that is, the long-tailed distributions satisfying the property $\sup_x G(x/2)/G(x) < \infty$.

Let G be an arbitrary distribution on \mathbf{R} with support unbounded from above and with finite mean value. For any $t \in (0, \infty]$, define the distribution G_t on \mathbf{R}^+ with the distribution tail

$$(1.1) \quad G_t(x) \equiv \min \left(1, \int_x^{x+t} G(u) du \right), \quad x > 0.$$

Note that any long-tailed distribution G satisfies the following relation, for any fixed $s > 0$:

$$(1.2) \quad G_s(x) = o(G_t(x)) \quad \text{as } t, x \rightarrow \infty.$$

We say (see [6]) that the subexponential distribution G on \mathbf{R}^+ is *strongly subexponential* (and write $G \in \mathcal{S}_*$) if the convolution of the distribution G_t with itself satisfies the equivalence $G_t * G_t(x) \sim 2G_t(x)$ as $x \rightarrow \infty$ uniformly in $t \in [1, \infty]$.

Criteria for strong subexponentiality are given in [6]. In particular, the class \mathcal{S}_* includes the following distributions:

- (i) the upper power distributions and, in particular, all distributions with regularly-varying-at-infinity tails;
- (ii) the lognormal distribution with the density $\exp\{-(\log x - \log \alpha)^2 / 2\sigma^2\} / x\sigma\sqrt{2\pi}$, $x > 0$, where $\sigma, \alpha > 0$;
- (iii) the Weibull distribution with the tail $G([x, \infty)) = e^{-x^\alpha}$, $x \geq 0$, where $\alpha \in (0, 1)$.

In Part 1 (see [1]) we assumed that X was a chain possessing an invariant measure π , that is, a measure solving the equation

$$(1.3) \quad \pi(\cdot) = \int_{\mathbf{R}} \pi(du) P(u, \cdot), \quad \pi(\mathbf{R}) = 1.$$

We have proved the following result about the asymptotic behavior of the tail $\pi(x)$ of the invariant measure π .

THEOREM 1. *Let X be an asymptotically homogeneous-in-space Markov chain such that the family (with respect to u) of jumps $\{\xi(u)\}$ is uniformly integrable. Let $m = \mathbf{E}\xi < 0$ and the distribution F of the random variable ξ be such that the distribution F_∞ (for a definition, see (1.1)) is upper power. If, for some bounded function $c(u)$, the convergence*

$$\frac{\mathbf{P}\{\xi(u) \geq t\}}{\mathbf{P}\{\xi \geq t\}} \longrightarrow c(u) \quad \text{as } t \rightarrow \infty$$

holds uniformly in u , then

$$\pi(x) \sim \frac{F_\infty(x)}{|m|} \int_{\mathbf{R}} c(u) \pi(du) \quad \text{as } x \rightarrow \infty.$$

Remark 1. Unfortunately, the boundedness condition for the function $c(u)$ was missed in the statement of Theorem 6 in [1].

In this third part we investigate the asymptotic behavior of the tail $\pi_n(x) = \mathbf{P}\{X(n) \geq x\}$ of the distribution π_n of the random variable $X(n)$ as $x \rightarrow \infty$ for fixed values of the time parameter n and for unboundedly growing n as well.

Let $\{\xi_k\}$ be a tuple of independent copies of ξ . Put $S_0 = 0$, $S_k = \xi_1 + \dots + \xi_k$, and $M_n = \max_{0 \leq k \leq n} S_k$. It is known (see, for example, [5, Chap. VI, section 9]) that the distribution of the homogeneous-in-space (see [1]) Markov chain $X(n) = (X(n + 1) + \xi_n)^+$ with zero initial state $X(0) = 0$ coincides with the distribution of M_n , that is,

$$(1.4) \quad \mathbf{P}\{X(0, n) \geq x\} = \mathbf{P}\{M_n \geq x\}.$$

Below we need the following theorem proved in [6].

THEOREM 2. *Let $m < 0$ and let the distribution of the random variable $\xi \mathbf{I}\{\xi \geq 0\}$ be strongly subexponential. Then*

$$\mathbf{P}\{M_n \geq x\} \sim \frac{F_{n|m}(x)}{|m|}$$

as $x \rightarrow \infty$ uniformly in $n \geq 1$.

As was mentioned above, one of the main topics in this paper is the investigation of Markov chains with asymptotically homogeneous drift. For such chains we obtain in section 2 the lower bound for the probability $\pi_n(x) = \mathbf{P}\{X(n) \geq x\}$. In section 3, the upper bounds are given for this probability. Combining these results in section 4 we get the theorem on the large deviation asymptotics for the asymptotically homogeneous-in-space Markov chain X .

2. Lower bound for large deviation probabilities for a prestationary chain. In this section we estimate from below the probability $\pi_n(x)$ for large values of n and x . This estimate is asymptotically correct. We start with some auxiliary results.

2.1. SLLN-type statements for a Markov sequence. Consider a nonhomogeneous in time Markov chain $Y = \{Y_n\}$. The initial distribution of this chain is assumed to be arbitrary. Let $\eta_{n+1}(u)$ be a random variable corresponding to the jump of the chain Y at time n from the state u , i.e.,

$$\mathbf{P}\{Y_{n+1} \in \cdot \mid Y_n = u\} = \mathbf{P}\{u + \eta_{n+1}(u) \in \cdot\}.$$

LEMMA 1. *Let the drift of the Markov chain Y_n be bounded from below by \hat{a} : $\mathbf{E}\eta_n(u) \geq \hat{a}$ for any time $n \geq 1$ and any state $u \in \mathbf{R}$. Moreover, let the family of random variables $\{|\eta_n(u)|, n \geq 1, u \in \mathbf{R}\}$ admit an integrable majorant, that is, there exists a random variable η with finite mean value such that $|\eta_n(u)| \leq_{\text{st}} \eta$ for each n and u . Then, for any initial distribution Y_0 ,*

$$\liminf_{n \rightarrow \infty} \frac{Y_n - Y_0}{n} \geq \hat{a} \quad \text{a.s.}$$

Proof. Fix $A > 0$. Define a threshold of the jump $\eta_n(u)$ at the level An as follows:

$$\eta_n^{[An]}(u) \equiv \eta_n(u) \mathbf{I}\{|\eta_n(u)| < An\}.$$

Let us consider a nonhomogeneous in time Markov chain Z_n , $Z_0 = Y_0$, with jumps $\eta_n^{[An]}(u)$:

$$\mathbf{P}\{Z_{n+1} \in \cdot \mid Z_n = u\} = \mathbf{P}\left\{u + \eta_n^{[An]}(u) \in \cdot\right\}.$$

By the construction of Z_n , we may estimate the probability of the event that the trajectories of Z_n and Y_n are different in the following way:

$$\begin{aligned} \mathbf{P}\left\{\sup_n |Z_n - Y_n| \neq 0\right\} &\leq \sum_{n=0}^{\infty} \mathbf{P}\{|Y_{n+1} - Y_n| \geq An\} \\ (2.1) \qquad \qquad \qquad &\leq \sum_{n=1}^{\infty} \mathbf{P}\{|\eta| \geq An\} \leq \mathbf{E} \frac{|\eta|}{A}. \end{aligned}$$

Put $\Delta_n^0 = \mathbf{E}\{Z_n - Z_{n-1} \mid Z_{n-1}\}$ and $\Delta_n^1 = Z_n - Z_{n-1} - \Delta_n^0$. Then

$$Z_n - Z_0 = \sum_{k=1}^n \Delta_k^0 + \sum_{k=1}^n \Delta_k^1 \equiv Z_n^0 + Z_n^1.$$

For every u we have the following equality and inequality:

$$\begin{aligned} \mathbf{E}\{\Delta_n^0 \mid Z_{n-1} = u\} &= \mathbf{E}\eta_n^{[An]}(u) = \mathbf{E}\eta_n(u) - \mathbf{E}\left\{\eta_n(u); |\eta_n(u)| \geq An\right\} \\ &\geq \hat{a} - \mathbf{E}\{|\eta|; |\eta| \geq An\}, \end{aligned}$$

in view of the conditions on the jumps $\eta_n(u)$. Therefore,

$$\frac{1}{n} \sum_{k=1}^n \Delta_k^0 \geq \hat{a} - \frac{1}{n} \sum_{k=1}^n \mathbf{E}\{|\eta|; |\eta| \geq Ak\}.$$

In view of the existence of $\mathbf{E}|\eta|$, the last inequality implies that uniformly in all elementary events ω ,

$$(2.2) \qquad \qquad \qquad \liminf_{n \rightarrow \infty} \frac{Z_n^0(\omega)}{n} \geq \hat{a}.$$

Since $\mathbf{E}\{\Delta_n^1 \mid Z_n\} = 0$, the process Z_n^1 is a martingale with respect to σ -fields $\sigma(Z_0, \dots, Z_{n-1})$. Let us prove that the increments of this martingale satisfy the condition

$$(2.3) \qquad \qquad \qquad \sum_{n=1}^{\infty} \frac{\mathbf{E}(\Delta_n^1)^2}{n^2} < \infty.$$

For every u , by the construction of Δ_n^1 and in view of the lemma conditions, we have

$$\begin{aligned} \mathbf{E}\{(\Delta_n^1)^2 \mid Z_{n-1} = u\} &= \mathbf{E}\left(\eta_n^{[An]}(u) - \mathbf{E}\eta_n^{[An]}(u)\right)^2 \\ &\leq \mathbf{E}\left(\eta_n^{[An]}(u)\right)^2 \leq \mathbf{E}\{\eta^2; |\eta| < An\}. \end{aligned}$$

Thus,

$$\sum_{n=1}^{\infty} \frac{\mathbf{E}(\Delta_n^1)^2}{n^2} \leq \sum_{n=1}^{\infty} \frac{\mathbf{E}\{\eta^2; |\eta| < An\}}{n^2}.$$

The last series converges for any A , since

$$\begin{aligned} \sum_{n=1}^{\infty} \frac{\mathbf{E}\{\eta^2; |\eta| < An\}}{n^2} &= \sum_{n=1}^{\infty} \frac{A^2}{n^2} \mathbf{E}\left\{\left(\frac{\eta}{A}\right)^2; \left|\frac{\eta}{A}\right| < n\right\} \\ &\leq \sum_{n=1}^{\infty} \frac{A^2}{n^2} \sum_{k=1}^n k^2 \mathbf{P}\left\{k-1 \leq \left|\frac{\eta}{A}\right| < k\right\} \\ &= A^2 \sum_{k=1}^{\infty} k^2 \mathbf{P}\left\{k-1 \leq \left|\frac{\eta}{A}\right| < k\right\} \sum_{n=k}^{\infty} \frac{1}{n^2} < \infty, \end{aligned}$$

by virtue of the equivalence $\sum_{n=k}^{\infty} 1/n^2 \sim 1/k$ and the existence of $\mathbf{E}|\eta|$.

So, the martingale Z_n^1 really satisfies condition (2.3) and we may apply Corollary 2 from [7, p. 534]. By this corollary, the following SLLN is valid for this martingale:

$$(2.4) \quad \lim_{n \rightarrow \infty} \frac{Z_n^1}{n} = 0 \quad \text{a.s.}$$

Relations (2.2) and (2.4) imply the inequality $\liminf_{n \rightarrow \infty} (Z_n - Z_0)/n \geq \hat{a}$ almost surely. Now the assertion of the lemma follows from (2.1) in view of the arbitrary choice of A .

LEMMA 2. *Let the drift of the Markov chain Y_n above some space level U be bounded below by $\hat{a} < 0$: $\mathbf{E}\eta_n(u) \geq \hat{a}$ for every $n \geq 1$ and $u \geq U$. Moreover, let the family of the random variables $\{|\eta_n(u)|, n \geq 1, u \geq U\}$ admit an integrable majorant; that is, there exists a random variable η with finite mean value such that $|\eta_n(u)| \leq_{\text{st}} \eta$ for every n and $u \geq U$. Then, for each $\varepsilon > 0$, the following convergence holds:*

$$\mathbf{P}\left\{Y_k \geq u - n(|\hat{a}| + \varepsilon) \text{ for all } k \leq n \mid Y_0 = u\right\} \longrightarrow 1$$

as $n \rightarrow \infty$ uniformly in $u \geq U + n(|\hat{a}| + \varepsilon)$.

Proof. It is sufficient to consider a new, also nonhomogeneous in time, Markov chain \tilde{Y}_n with jumps $\tilde{\eta}_n(u)$, where $\tilde{\eta}_n(u)$ coincides with $\eta_n(u)$ if $u \geq U$, and $\tilde{\eta}_n(u)$ is equal to \hat{a} if $u < U$. Applying Lemma 1 we complete the proof.

2.2. Lower bound for the large deviation probabilities. The initial distribution π_0 of the chain X is assumed to be arbitrary. Let f be a positive nonincreasing long-tailed function. Let the function $\underline{c}(u) \geq 0$ be such that, for every $U > 0$,

$$\frac{\mathbf{P}\{\xi(u) \geq x\}}{f(x)} \geq \underline{c}(u) + o(1)$$

as $x \rightarrow \infty$ uniformly in $|u| \leq U$. Denote

$$c_{\infty} \equiv \lim_{U \rightarrow \infty} \liminf_{n \rightarrow \infty} \int_{-U}^U \underline{c}(u) \pi_n(du) \in [0, \infty].$$

In particular, if the distribution π_n converges in the total variation norm to some (invariant with necessity) measure π , i.e., if the convergence

$$(2.5) \quad \sup_B |\pi_n(B) - \pi(B)| \longrightarrow 0$$

holds as $n \rightarrow \infty$, where the supremum is taken over all Borel sets on the real line, then $c_{\infty} = \int_{\mathbf{R}} \underline{c}(u) \pi(du)$. If the function $\underline{c}(u)$ is continuous, the last equality remains true in the case of weak convergence $\pi_n \Rightarrow \pi$.

Put

$$\underline{a} \equiv \liminf_{u \rightarrow \infty} \mathbf{E} \xi(u).$$

It was established in Lemma 1 in [2] that the existence of the invariant measure together with the condition $\sup_u \mathbf{E} |\xi(u)| < \infty$ implies $\underline{a} \leq 0$. Note that in the present section we do not put any restrictions on the sign of \underline{a} .

The following lemma is valid.

LEMMA 3. Put $a = -\underline{a}$ if $\underline{a} < 0$, and let a be an arbitrary positive real number otherwise. Let there exist a level U such that the family $\{|\xi(u)|, u \geq U\}$ admits an integrable majorant. Then the following estimate holds:

$$\liminf_{n, x \rightarrow \infty} \frac{\pi_n(x)}{\int_x^{x+na} f(y) dy} \geq \frac{c_\infty}{a}.$$

Proof. It is sufficient to consider the case $c_\infty > 0$ only. Let $c' < c'' < c_\infty$. The definitions of $\underline{c}(u)$ and c_∞ imply the existence of $U' > 0$ such that

$$\liminf_{n \rightarrow \infty} \int_{-U'}^{U'} \frac{\mathbf{P}\{\xi(u) \geq x\}}{f(x)} \pi_n(du) \geq c''$$

uniformly in all sufficiently large x . Since the function $f(x)$ is long-tailed and non-increasing, $f(x - u)/f(x) \rightarrow 1$ as $x \rightarrow \infty$ uniformly in $|u| \leq U'$. Therefore, there exist N and x_0 such that

$$(2.6) \quad \int_{-U'}^{U'} \frac{\mathbf{P}\{u + \xi(u) \geq x\}}{f(x)} \pi_n(du) \geq c'$$

uniformly in $n \geq N$ and $x \geq x_0$.

Consider the event $A_{i,n} \equiv A_{i,n}(x)$, $i \in [1, n]$, which occurs if $X(i - 1) < x$ and $X(j) \geq x$ for any $j \in [i, n]$. First, the events $A_{i,n}$, $i \in [1, n]$, are disjoint. Second, $\cup_{i=1}^n A_{i,n} = \{X(n) \geq x\}$. Thus

$$(2.7) \quad \pi_n(x) = \sum_{i=1}^n \mathbf{P}\{A_{i,n}\}.$$

For $v \in [x, \infty)$, introduce the probability $p_i(v)$ by the equality

$$p_i(v) = \mathbf{P}\{X(v, j) \geq x \text{ for any } j \leq i\}.$$

Fix $\varepsilon > 0$ and put $b = a + \varepsilon$. Since the event $A_{i,n}(U') \equiv \{X(i - 1) \in [-U', U'], X(i) \geq x + (n - i)b, \text{ and } X(j) \geq x \text{ for any } j \in [i + 1, n]\}$ implies the event $A_{i,n}$, the following inequality holds:

$$\mathbf{P}\{A_{i,n}\} \geq \int_{-U'}^{U'} \pi_{i-1}(du) \int_{x+(n-i)b}^\infty \mathbf{P}\{u + \xi(u) \in dv\} p_{n-i}(v).$$

Therefore,

$$(2.8) \quad \mathbf{P}\{A_{i,n}\} \geq \int_{-U'}^{U'} \pi_{i-1}(du) \mathbf{P}\{u + \xi(u) \geq x + (n - i)b\} \min_{v \geq x+(n-i)b} p_{n-i}(v).$$

By the definition of a and b , we have the inequality $\mathbf{E}\xi(v) \geq -b + \varepsilon/2$ for all sufficiently large v . Thus, by virtue of Lemma 2,

$$\min_{v \geq x+(n-i)b} p_{n-i}(v) \rightarrow 1 \quad \text{as } x \text{ and } n-i \rightarrow \infty.$$

Substituting this convergence into (2.8), we obtain the inequality

$$\liminf_{n-i, x \rightarrow \infty} \frac{\mathbf{P}\{A_{i,n}\}}{f(x+(n-i)b)} \geq \liminf_{n-i, x \rightarrow \infty} \int_{-U'}^{U'} \frac{\mathbf{P}\{u + \xi(u) \geq x + (n-i)b\}}{f(x+(n-i)b)} \pi_{i-1}(du) \geq c'$$

in view of (2.6). Using the last inequality, we derive from (2.7) the estimate

$$\pi_n(x) \geq (c' - \varepsilon) \sum_{i=N}^{n-I} f(x+(n-i)b),$$

which is valid for arbitrary slow growing I and for all sufficiently large x . Since the function f is long-tailed and nonincreasing, for every fixed I ,

$$\sum_{i=N}^{n-I} f(x+(n-i)b) \sim \frac{1}{b} \int_x^{x+nb} f(y) dy \quad \text{as } x \rightarrow \infty.$$

Therefore,

$$\liminf_{n, x \rightarrow \infty} \frac{\pi_n(x)}{\int_x^{x+nb} f(y) dy} \geq \frac{c' - \varepsilon}{b}.$$

Since $b = a + \varepsilon$, $c' < c_\infty$, and $\varepsilon > 0$ were chosen arbitrarily, the lemma proof is complete.

3. Upper bounds for the tails of prestationary and stationary distributions.

3.1. Upper bound with a “nonexact” multiple constant. The following lemma generalizes Lemma 2 from [1] in the part which is related to subexponential distributions.

LEMMA 4. *Let ζ be an unbounded-from-above random variable with distribution G and with negative mean $\mathbf{E}\zeta < 0$. Let there exist a level U such that, for each $t \geq U$,*

$$(3.1) \quad \xi(u) \leq_{\text{st}} \zeta \quad \text{if } u \geq U,$$

$$(3.2) \quad \mathbf{P}\{u + \xi(u) \geq t\} \leq G(t) \quad \text{if } u < U.$$

Then the following assertions are true:

(i) *If the distribution G_∞ (for a definition, see (1.1)) is subexponential and the chain X admits (not unique, in general) an invariant measure π , then the following estimate holds:*

$$(3.3) \quad \limsup_{x \rightarrow \infty} \frac{\pi(x)}{G_\infty(x)} \leq \frac{1}{|\mathbf{E}\zeta|};$$

(ii) *if the distribution G is strongly subexponential and the initial distribution $X(0)$ is bounded from above, then the following estimate holds:*

$$(3.4) \quad \limsup_{x \rightarrow \infty} \sup_{n \geq 1} \frac{\pi_n(x)}{G_n|\mathbf{E}\zeta|(x)} \leq \frac{1}{|\mathbf{E}\zeta|}.$$

Proof. Without loss of generality we assume that $X(0) \leq 0$ and $U = 0$. Consider the homogeneous-in-space Markov chain $\{Y_n\}$ with nonnegative values defined by the equalities $Y_{n+1} = (Y_n + \zeta_{n+1})^+$, where ζ_n are independent copies of ζ . By virtue of conditions (3.1) and (3.2), the Markov chain Y_n dominates the chain $X(n)$ and, therefore,

$$(3.5) \quad \pi(x) \leq \pi^Y(x),$$

where π^Y is the invariant measure of the chain Y .

Since the distribution G_∞ is subexponential, it follows from Theorem 2(B) in [9] that

$$\pi^Y(x) \sim \frac{G_\infty(x)}{|\mathbf{E}\zeta|} \quad \text{as } x \rightarrow \infty.$$

Combined with (3.5) this implies (3.3).

Inequality (3.4) also follows from the majorization of the chain $X(n)$ by the chain Y_n . By majorization, for every n and x we have the estimate $\pi_n(x) \leq \pi_n^Y(x) = \mathbf{P}\{Y_n \geq x\}$. It remains to apply Theorem 2. The lemma is proved.

3.2. Some auxiliary results. Let ζ_1, ζ_2, \dots be i.i.d. random variables.

LEMMA 5. *Let $\mathbf{E}\zeta_1 < 0$. Then the series*

$$\sum_{n=1}^{\infty} \mathbf{P}\{\zeta_1 + \dots + \zeta_k \geq 0 \text{ for all } k \leq n\}$$

converges.

Proof. Consider the homogeneous-in-space Markov chain defined by the equalities $Y_{n+1} = (Y_n + \zeta_{n+1})^+$ with the initial value $Y_0 = 1$. We have the inequality and the equality $\mathbf{P}\{\zeta_1 + \dots + \zeta_k \geq 0 \text{ for all } k \leq n\} \leq \mathbf{P}\{Y_k > 0 \text{ for all } k \leq n | Y_0 = 1\} = \mathbf{P}\{\eta > n\}$, where η is the first hitting time of the state 0 by the chain $\{Y_n\}$. Since $\mathbf{E}\zeta_1 < 0$, the chain $\{Y_n\}$ is positive recurrent, that is, $\mathbf{E}\eta < \infty$. Thus, the series $\sum_{n=1}^{\infty} \mathbf{P}\{\eta > n\}$ converges and the lemma is proved.

In the following theorem, the local behavior of the function is studied which is dominated by the long-tailed function. Let a positive nonincreasing integrable function $f(x)$ be long-tailed. Since the function f does not increase, it is long-tailed if and only if there exists a sequence $\Delta(x) \rightarrow \infty$ such that

$$\frac{f(x)}{f(x - \Delta(x))} \rightarrow 1 \quad \text{as } x \rightarrow \infty.$$

Let c be an arbitrary positive constant. Put

$$f_n(x) = \int_x^{x+cn} f(y) dy.$$

We have the equivalence

$$(3.6) \quad f_n(x - \Delta(x)) \sim f_n(x)$$

as $x \rightarrow \infty$ uniformly in all n . Let a nonnegative function $h_n(x)$ be such that, for each n , it is nonincreasing in x .

LEMMA 6. Let $h_n(x) \leq f_n(x)$ for any n and x . Then there exists a sequence of intervals $[x_n^-, x_n^+] \subseteq [x - \Delta(x), x]$ such that $x_n^+ - x_n^- \rightarrow \infty$ and $h_n(x_n^-) - h_n(x_n^+) = o(f_n(x))$ as $n, x \rightarrow \infty$.

Proof. Choose a sequence $u(x) \rightarrow \infty$ such that $u(x) = o(\Delta(x))$ as $x \rightarrow \infty$ and $l(x) = \Delta(x)/u(x)$ is a natural number. By the choice of $u(x)$ we have the convergence $l(x) \rightarrow \infty$ as $x \rightarrow \infty$.

By virtue of (3.6), it is sufficient to prove that, for any n and x , there exists a point $x_n^- \in [x - \Delta(x), x - u(x)]$ such that $h_n(x_n^-) - h_n(x_n^- + u(x)) = o(f_n(x_n^-))$ as $n, x \rightarrow \infty$ (and put $x_n^+ = x_n^- + u(x)$). We argue by the rule of contraries and assume that the last relation does not hold. Then there exist a number $\varepsilon > 0$, subsequences $n_k \rightarrow \infty$, $k \rightarrow \infty$, and $x_k \rightarrow \infty$, $k \rightarrow \infty$, such that, for any $y \in [x_k - \Delta(x_k), x_k - u(x_k)]$, the following inequality holds:

$$(3.7) \quad h_{n_k}(y) - h_{n_k}(y + u(x_k)) \geq \varepsilon f_{n_k}(y).$$

In particular, in view of the inequality $h_{n_k} \leq f_{n_k}$,

$$h_{n_k}(y + u(x_k)) \leq h_{n_k}(y) - \varepsilon f_{n_k}(y) \leq (1 - \varepsilon) h_{n_k}(y).$$

Therefore,

$$\begin{aligned} h_{n_k}(x_k - u(x_k)) &\leq (1 - \varepsilon)^{l(x_k)-1} h_{n_k}(x_k - \Delta(x_k)) \\ &\leq (1 - \varepsilon)^{l(x_k)-1} f_{n_k}(x_k - \Delta(x_k)) = o(f_{n_k}(x_k - u(x_k))) \end{aligned}$$

as $k \rightarrow \infty$ by virtue of $l(x_k) \rightarrow \infty$ and (3.6). It contradicts (3.7) with $y = x_k - u(x_k)$. This contradiction proves the lemma.

3.3. Upper bound with an “exact” multiple constant. Here we assume that the initial distribution π_0 is concentrated on the set bounded from above.

LEMMA 7. Let an unbounded-from-above random variable ζ with distribution G be such that the distribution of the random variable $\zeta \mathbf{I}\{\zeta \geq 0\}$ is strongly subexponential and $\mathbf{E}\zeta < 0$. Let there exist a level U such that (3.1) holds. Moreover, let, for some function $c(v) \leq 1$, the inequality

$$(3.8) \quad \mathbf{P}\{\xi(u) \geq t\} \leq c(u) G(t)$$

hold for $u \geq U, t \geq U$ and

$$(3.9) \quad \mathbf{P}\{u + \xi(u) \geq t\} \leq c(u) G(t)$$

for $u < U, t \geq U$. If, in addition, convergence in total variation (2.5) holds, then

$$(3.10) \quad \pi_n(x) \leq (1 + o(1)) \sum_{k=1}^n c_{k-1} G(x + (n - k) |\mathbf{E}\zeta|),$$

as $x \rightarrow \infty$ uniformly in all $n \geq 1$, where $c_k \equiv \int_{\mathbf{R}} c(u) \pi_k(du) \geq 0$. In particular,

$$(3.11) \quad \limsup_{n, x \rightarrow \infty} \frac{\pi_n(x)}{G_n|\mathbf{E}\zeta|(x)} \leq \frac{c_\infty}{|\mathbf{E}\zeta|},$$

where $c_\infty \equiv \lim_{k \rightarrow \infty} c_k = \int_{\mathbf{R}} c(u) \pi(du) \geq 0$.

Proof. Since the time parameter n takes its values in the countable set, the distribution G is long-tailed, and $c_n \rightarrow c_\infty$, it is sufficient to check the following two relations: For any *fixed* n ,

$$(3.12) \quad \pi_n(x) \leq (1 + o(1)) G(x) \sum_{k=1}^n c_{k-1} \quad \text{as } x \rightarrow \infty$$

and (3.11). Let us prove the first relation by induction. For every $U' \in (U, x)$ we have the equality

$$\pi_{n+1}(x) = \left(\int_{-\infty}^{U'} + \int_{U'}^{x-U'} + \int_{x-U'}^{\infty} \right) \mathbf{P}\{u + \xi(u) \geq x\} \pi_n(du) \equiv I_1 + I_2 + I_3.$$

In view of conditions (3.8) and (3.9), for any fixed U' , we have the estimate

$$\limsup_{x \rightarrow \infty} \frac{I_1}{G(x)} \leq \int_{-\infty}^{U'} c(u) \pi_n(du) \leq c_n.$$

In view of $c(v) \leq 1$ and condition (3.8), the second term I_2 admits the following estimate:

$$I_2 \leq \int_{U'}^{x-U'} G(x-u) \pi_n(du).$$

Integrating by parts, we arrive at the inequality

$$\begin{aligned} I_2 &\leq -G(x-u) \pi_n(u) \Big|_{U'}^{x-U'} + \int_{U'}^{x-U'} \pi_n(x-u) G(du) \\ &\leq G(x-U') \pi_n(U') + \int_{U'}^{x-U'} \pi_n(x-u) G(du). \end{aligned}$$

Using the inductive hypothesis, we get the estimate

$$\limsup_{x \rightarrow \infty} \frac{I_2}{G(x)} \leq cG(U') + c \limsup_{x \rightarrow \infty} \int_{U'}^{x-U'} \frac{G(x-u)}{G(x)} G(du).$$

Since the distribution G is subexponential, it follows from relation (2) in [4] that the value of \limsup on the right-hand side of the last inequality may be made arbitrarily small by the appropriate choice of U' . Thus, $\lim_{U' \rightarrow \infty} \limsup_{x \rightarrow \infty} I_2/G(x) = 0$.

The third term I_3 does not exceed $\pi_n(x-U')$. Thus, by the induction assumption and by the fact that the distribution G is long-tailed, the estimate

$$\limsup_{x \rightarrow \infty} \frac{I_3}{G(x)} \leq \sum_{k=1}^n c_{k-1}$$

is valid for any fixed U' . Combining the estimates for I_1 , I_2 , and I_3 , we deduce the induction step $n \rightarrow n+1$.

Now let us prove relation (3.11). Without loss of generality, assume $X(0) \leq U$. Choose the sequence of points x_k , $k = 1, 2, \dots$, such that $x_{k+1} - x_k \rightarrow \infty$ and

$G(x_{k+1}) \sim G(x_k)$ as $k \rightarrow \infty$. In particular, $G_{n|\mathbf{E}\zeta|}(x_{k+1}) \sim G_{n|\mathbf{E}\zeta|}(x_k)$ as $k \rightarrow \infty$ uniformly in all n . So, relation (3.11) is equivalent to the relation

$$(3.13) \quad \limsup_{n,k \rightarrow \infty} \frac{\pi_n(x_k)}{G_{n|\mathbf{E}\zeta|}(x_k)} \leq \frac{c_\infty}{|\mathbf{E}\zeta|}.$$

By virtue of Theorem 4, for any fixed i , the function $h_n(x) \equiv \pi_{n-i}(x)$ satisfies the conditions of Lemma 6 with $f(x) \equiv \text{const} \cdot G(x)$ and $c = |\mathbf{E}\zeta|$. Hence, there exist sequences x_{kn}^- and x_{kn}^+ such that $[x_{kn}^-, x_{kn}^+] \subseteq [x_k, x_{k+1}]$, $x_{kn}^+ - x_{kn}^- \rightarrow \infty$, and $\pi_{n-i}(x_{kn}^-) - \pi_{n-i}(x_{kn}^+) = o(G_{n|\mathbf{E}\zeta|}(x_{k+1}))$ as $n, k \rightarrow \infty$. Therefore, without loss of generality, for convenience, we may assume that there exists (for any fixed i) the function x_n^- such that $x_n^- \in [0, x]$, $x - x_n^- \rightarrow \infty$, and

$$(3.14) \quad \pi_{n-i}([x_n^-, x_n^+]) = \pi_{n-i}(x_n^-) - \pi_{n-i}(x) = o(G_{n|\mathbf{E}\zeta|}(x)), \quad n, x \rightarrow \infty.$$

Condition (3.9) implies (3.2). Consider the homogeneous Markov chain $\{Y_n\}$ with nonzero states defined in the proof of Lemma 4. Since the chain $\{U + Y_n\}$ dominates the chain $\{X(n)\}$, given $X(0) \leq U + Y_0$, the chains $\{X(n)\}$ and $\{U + Y_n\}$ can be constructed on the same probability space in such a way that

$$(3.15) \quad X(n) \leq U + Y_n$$

for any n with probability 1. All characteristics of the chain $\{X(n)\}$ will be completed by an upper index X and the characteristics of the chain $\{U + Y_n\}$ by an upper index Y .

The events $A_{i,n}$ and the probabilities $p_i(v)$ were defined in the proof of Lemma 3. Let us turn again to equality (2.7). For any fixed i , estimates (3.4) and (1.2) imply the following relations:

$$\mathbf{P}\{A_{i,n}^X\} \leq \mathbf{P}\{X(i) \geq x\} = \pi_i(x) = O(G_{i|\mathbf{E}\zeta|}(x)) = o(G_{n|\mathbf{E}\zeta|}(x))$$

as $n, x \rightarrow \infty$. Therefore, there exists an unboundedly growing sequence $I = I(n, x)$ such that

$$(3.16) \quad \pi_n(x) = \sum_{i=I}^n \mathbf{P}\{A_{i,n}^X\} + o(G_{n|\mathbf{E}\zeta|}(x)) \quad \text{as } n, x \rightarrow \infty.$$

It turns out that any finite number of the last summands in this sum has order $o(G_{n|\mathbf{E}\zeta|}(x))$. This more delicate observation will be checked at the end of the proof.

For the probability of the event $A_{i,n}^X$, we have the equality

$$\mathbf{P}\{A_{i,n}^X\} = \int_{-\infty}^x \pi_{i-1}^X(du) \int_x^\infty P^X(u, dv) p_{n-i}^X(v).$$

Using (3.15) with $X(0) = U + Y_0 = v$, we obtain $p_{n-i}^X(v) \leq p_{n-i}^Y(v)$. Hence,

$$(3.17) \quad \mathbf{P}\{A_{i,n}^X\} \leq \int_{-\infty}^x \pi_{i-1}^X(du) \int_x^\infty P^X(u, dv) p_{n-i}^Y(v).$$

Being space homogeneous, the chain $U + Y_n$ is stochastically increasing. Thus, the function $p_{n-i}^Y(v)$ does not decrease in v and, therefore, is a function of bounded variation. Integrating by parts, we may rewrite the internal integral in the following way:

$$\int_x^\infty P^X(u, dv) p_{n-i}^Y(v) = P^X(u, [x, \infty)) p_{n-i}^Y(x) + \int_x^\infty P^X(u, [v, \infty)) dp_{n-i}^Y(v).$$

Estimating $P^X(u, [v, \infty))$ on the right-hand side of the last inequality via condition (3.9) and integrating back by parts, we obtain recursively the estimate and the equality, for any $u \in (-\infty, U)$,

$$\begin{aligned} \int_x^\infty P^X(u, dv) p_{n-i}^Y(v) &\leq c(u) \left(G(x) p_{n-i}^Y(x) + \int_x^\infty G(v) dp_{n-i}^Y(v) \right) \\ (3.18) \qquad \qquad \qquad &= c(u) \int_x^\infty G(dv) p_{n-i}^Y(v). \end{aligned}$$

In the same manner, in view of condition (3.8) and inequality $c(u) \leq 1$ we obtain, for any $u \in [U, x)$,

$$(3.19) \qquad \int_x^\infty P^X(u, dv) p_{n-i}^Y(v) \leq c(u) \int_x^\infty G(dv - u) p_{n-i}^Y(v)$$

$$(3.20) \qquad \qquad \qquad \leq \int_x^\infty G(dv - u) p_{n-i}^Y(v).$$

Since the function $G(t)$ is long-tailed, there exists an unboundedly growing (as $x \rightarrow \infty$) level V such that $G(v - u) \sim G(v)$ as $u \in [U, V)$ and $v \geq x$. Then inequality (3.19), for $u \in [U, V)$, can be rewritten as

$$(3.21) \qquad \int_x^\infty P^X(u, dv) p_{n-i}^Y(v) \leq (c(u) + o(1)) \int_x^\infty G(dv) p_{n-i}^Y(v), \quad x \rightarrow \infty.$$

Substituting (3.18), (3.20), and (3.21) into (3.17), we arrive at the inequality

$$\begin{aligned} \mathbf{P}\{A_{i,n}^X\} &\leq (1 + o(1)) \int_{-\infty}^V \pi_{i-1}^X(du) c(u) \int_x^\infty G(dv) p_{n-i}^Y(v) \\ &\quad + \int_V^x \pi_{i-1}^X(du) \int_x^\infty G(dv - u) p_{n-i}^Y(v). \end{aligned}$$

Recalling the definition of the constants c_i , we obtain

$$\mathbf{P}\{A_{i,n}^X\} \leq (c_{i-1} + o(1)) \int_x^\infty G(dv) p_{n-i}^Y(v) + \int_V^x \pi_{i-1}^X(du) \int_x^\infty G(dv - u) p_{n-i}^Y(v)$$

as $x \rightarrow \infty$ uniformly in $i \in [1, n]$. Summing up these inequalities with respect to i from $I(n, x)$ to n and taking into account the convergence $c_i \rightarrow c_\infty$ as $i \rightarrow \infty$, we deduce from (3.16) the relation, as $n, x \rightarrow \infty$,

$$\begin{aligned} \pi_n^X(x) &\leq (c_\infty + o(1)) \int_x^\infty G(dv) \sum_{i=1}^n p_{n-i}^Y(v) \\ (3.22) \qquad &+ \sum_{i=I}^n \int_V^x \pi_{i-1}^X(du) \int_x^\infty G(dv - u) p_{n-i}^Y(v) + o(G_{n|\mathbf{E}\zeta}(x)). \end{aligned}$$

For the homogeneous-in-space Markov chain $U + Y$, we have the inequality

$$\begin{aligned} \mathbf{P}\{A_{i,n}^Y\} &\geq \int_U^V \pi_{i-1}^Y(du) \int_x^\infty P^Y(u, dv) p_{n-i}^Y(v) \\ &= \int_U^V \pi_{i-1}^Y(du) \int_x^\infty G(dv - u) p_{n-i}^Y(v). \end{aligned}$$

By virtue of the choice of the level V , we have

$$\mathbf{P}\{A_{i,n}^Y\} \geq \int_U^V \pi_{i-1}^Y(du) \int_x^\infty G(dv) p_{n-i}^Y(v) = \pi_{i-1}^Y([U, V)) \int_x^\infty G(dv) p_{n-i}^Y(v).$$

Since $V \rightarrow \infty$,

$$\mathbf{P}\{A_{i,n}^Y\} \geq (1 + o(1)) \int_x^\infty G(dv) p_{n-i}^Y(v)$$

as $x \rightarrow \infty$ uniformly in $i \in [1, n]$. Summing up these inequalities with respect to i from 1 to n , we deduce from (2.7) the relation

$$\pi_n^Y(x) = \sum_{i=1}^n \mathbf{P}\{A_{i,n}^Y\} \geq (1 + o(1)) \int_x^\infty G(dv) \sum_{i=1}^n p_{n-i}^Y(v).$$

It follows from (1.4) and Theorem 2 that, as $x \rightarrow \infty$ uniformly in $n \geq 1$,

$$\pi_n^Y(x) \leq \frac{1 + o(1)}{|\mathbf{E}\zeta|} G_{n|\mathbf{E}\zeta|}(x).$$

The last two inequalities imply the estimate

$$\int_x^\infty G(dv) \sum_{i=1}^n p_{n-i}^Y(v) \leq \frac{1 + o(1)}{|\mathbf{E}\zeta|} G_{n|\mathbf{E}\zeta|}(x) \quad \text{as } x \rightarrow \infty.$$

Substituting it into (3.22), we arrive at the relation

$$(3.23) \quad \pi_n^X(x) \leq \frac{c_\infty + o(1)}{|\mathbf{E}\zeta|} G_{n|\mathbf{E}\zeta|}(x) + \sum_{i=1}^n \int_V^x \pi_{i-1}^X(du) q_{n-i}(u)$$

as $n, x \rightarrow \infty$, where $q_{n-i}(u) = \int_x^\infty G(dv - u) p_{n-i}^Y(v)$.

The previous considerations also imply the following estimate for the chain $U + Y$, which is valid as $x \rightarrow \infty$ uniformly in $n \geq 1$:

$$(3.24) \quad \sum_{i=1}^n \int_V^x \pi_{i-1}^Y(du) \int_x^\infty G(dv - u) p_{n-i}^Y(v) = o(G_{n|\mathbf{E}\zeta|}(x)).$$

It remains to estimate the general term of the sum in (3.23). Integration by parts implies the equality

$$\int_V^x \pi_{i-1}^X(du) q_{n-i}(u) = -\pi_{i-1}^X(u) q_{n-i}(u) \Big|_V^x + \int_V^x \pi_{i-1}^X(u) dq_{n-i}(u).$$

In view of Lemma 4 and Theorem 2, $\pi_{i-1}^X(u) \leq c\pi_{i-1}^Y(u)$ for some $c < \infty$ and for all u and i . Hence,

$$\int_V^x \pi_{i-1}^X(du) q_{n-i}(u) \leq c\pi_{i-1}^Y(V) q_{n-i}(V) + c \int_V^x \pi_{i-1}^Y(u) dq_{n-i}(u).$$

Again integrating by parts, we obtain

$$\begin{aligned}
 & \int_V^x \pi_{i-1}^X(du) q_{n-i}(u) \\
 & \leq c\pi_{i-1}^Y(V) q_{n-i}(V) + c\pi_{i-1}^Y(u) q_{n-i}(u) \Big|_V^x + c \int_V^x \pi_{i-1}^Y(du) q_{n-i}(u) \\
 (3.25) \quad & \leq c\pi_{i-1}^Y(x) q_{n-i}(x) + c \int_V^x \pi_{i-1}^Y(du) q_{n-i}(u).
 \end{aligned}$$

In the same way, we obtain the estimate

$$\begin{aligned}
 \int_V^x \pi_{i-1}^X(du) q_{n-i}(u) & = \left(\int_V^{x_n^-} + \int_{x_n^-}^x \right) \pi_{i-1}^X(du) q_{n-i}(u) \leq c\pi_{i-1}^Y(x_n^-) q_{n-i}(x_n^-) \\
 (3.26) \quad & + c \int_V^{x_n^-} \pi_{i-1}^Y(du) q_{n-i}(u) + \pi_{i-1}^X([x_n^-, x]).
 \end{aligned}$$

Fix a natural number J . Applying estimate (3.25) if $i \in [I, n - J]$ and estimate (3.26) if $i \in [n - J + 1, n]$, we deduce the inequality

$$\begin{aligned}
 \sum_{i=I}^n \int_V^x \pi_{i-1}^X(du) q_{n-i}(u) & \leq c \sum_{i=I}^{n-J} \pi_{i-1}^Y(x) q_{n-i}(x) + c \sum_{i=n-J+1}^n \pi_{i-1}^Y(x_n^-) q_{n-i}(x_n^-) \\
 & + c \sum_{i=I}^n \int_V^x \pi_{i-1}^Y(du) q_{n-i}(u) + \sum_{i=n-J+1}^n \pi_{i-1}^X([x_n^-, x]).
 \end{aligned}$$

By virtue of (3.24), the third sum on the right-hand side of the estimate is of the order $o(G_{n|\mathbf{E}\zeta|}(x))$; the fourth sum is of the same order, for any fixed J , in view of (3.14). Therefore, for sufficiently slow-growing level $J \equiv J_n(x) \rightarrow \infty$, the following estimate holds:

$$\begin{aligned}
 & \sum_{i=I}^n \int_V^x \pi_{i-1}^X(du) q_{n-i}(u) \\
 & \leq c \sum_{i=I}^{n-J} \pi_{i-1}^Y(x) q_{n-i}(x) + c \sum_{i=n-J+1}^n \pi_{i-1}^Y(x_n^-) q_{n-i}(x_n^-) + o(G_{n|\mathbf{E}\zeta|}(x)) \\
 (3.27) \quad & \equiv c\Sigma_1 + c\Sigma_2 + o(G_{n|\mathbf{E}\zeta|}(x)) \quad \text{as } n, x \rightarrow \infty.
 \end{aligned}$$

Since

$$\begin{aligned}
 q_{n-i}(x) & = \int_x^\infty G(dv - x) p_{n-i}^Y(v) \\
 & = \int_x^\infty G(dv - x) \mathbf{P}\{Y_k \geq x \text{ for all } k \in [2, n - i + 1] \mid Y_1 = v\} \\
 & = \mathbf{P}\{Y_k \geq x \text{ for all } k \in [1, n - i + 1] \mid Y_0 = x\},
 \end{aligned}$$

then

$$\begin{aligned}
 \Sigma_1 & = \sum_{i=I}^{n-J} \pi_{i-1}^Y(x) \mathbf{P}\{\zeta_1 + \dots + \zeta_k \geq 0 \text{ for all } k \leq n - i + 1\} \\
 & = O(G_{n|\mathbf{E}\zeta|}(x)) \sum_{i=I}^{n-J} \mathbf{P}\{\zeta_1 + \dots + \zeta_k \geq 0 \text{ for all } k \leq n - i + 1\}.
 \end{aligned}$$

Since $J \rightarrow \infty$, it follows from Lemma 5 that $\sum_{i=I}^{n-J} \mathbf{P}\{\zeta_1 + \dots + \zeta_k \geq 0 \text{ for all } k \leq n - i + 1\} \rightarrow 0$. Therefore,

$$(3.28) \quad \Sigma_1 = o(G_{n|\mathbf{E}\zeta|}(x)) \quad \text{as } n, x \rightarrow \infty.$$

Now prove that

$$(3.29) \quad \Sigma_2 = o(G_{n|\mathbf{E}\zeta|}(x)) \quad \text{as } n, x \rightarrow \infty.$$

Since the sequence $J_n(x)$ may increase as slowly as possible, it is sufficient to check that, for any fixed i , $\pi_{n-i}^Y(x_n^-) q_{i-1}(x_n^-) = o(G_{n|\mathbf{E}\zeta|}(x))$ holds as $n, x \rightarrow \infty$. In view of (3.14) and Lemma 4, we have $\pi_{n-i}^Y(x_n^-) = \pi_{n-i}^Y(x) + o(G_{n|\mathbf{E}\zeta|}(x)) = O(G_{n|\mathbf{E}\zeta|}(x))$. Hence,

$$\begin{aligned} \pi_{n-i}^Y(x_n^-) q_{i-1}(x_n^-) &= O(G_{n|\mathbf{E}\zeta|}(x)) q_{i-1}(x_n^-) \leq O(G_{n|\mathbf{E}\zeta|}(x)) P(x_n^-, [x, \infty)) \\ &= o(G_{n|\mathbf{E}\zeta|}(x)), \end{aligned}$$

since $x - x_n^- \rightarrow \infty$ and $P(x_n^-, [x, \infty)) \rightarrow 0$. So, relation (3.29) is proved.

Substituting (3.28) and (3.29) into (3.27), we arrive at the following relation:

$$\sum_{i=I}^n \int_V^x \pi_{i-1}^X(du) q_{n-i}(u) = o(G_{n|\mathbf{E}\zeta|}(x)) \quad \text{as } n, x \rightarrow \infty.$$

Taking into account the last relation and (3.23), we arrive at estimate (3.11) and, simultaneously, at the conclusion of the lemma.

4. Uniform-in-time theorem on large deviation probabilities. Now we consider the asymptotically homogeneous-in-space Markov chain, that is, we assume the weak convergence $\xi(u) \Rightarrow \xi$ as $u \rightarrow \infty$. Let $\mathbf{E}\xi < 0$. In the present section we also assume that the support of the initial distribution π_0 is bounded from above and that convergence in total variation (2.5) holds.

Let ζ be an unbounded-from-above random variable with distribution G and negative mean value. Assume that the distribution of the random variable $\zeta \mathbf{I}\{\zeta \geq 0\}$ is strongly subexponential.

THEOREM 3. *Let there exist a level U such that the family of random variables $\{|\xi(u)|, u \geq U\}$ admits an integrable majorant. Let condition*

$$\mathbf{P}\{u + \xi(u) \geq t\} \leq G(t) \quad \text{if } u < U, \quad \xi(u) \leq_{\text{st}} \zeta \quad \text{if } u \geq U$$

hold for $t \geq U$. Suppose that, for some function $c(u) \leq 1$, the uniform-in- u convergence $\mathbf{P}\{\xi(u) \geq t\}/G(t) \rightarrow c(u)$ holds as $t \rightarrow \infty$. Then relation

$$\pi_n(x) = (1 + o(1)) \sum_{k=1}^n c_{k-1} G(x + (n - k) |\mathbf{E}\xi|)$$

holds as $x \rightarrow \infty$ uniformly in $n \geq 1$, where the constants c_k are defined by the equalities $c_k = \int_{\mathbf{R}} c(u) \pi_k(du)$. In particular,

$$\pi_n(x) = \left(\frac{c_\infty}{|\mathbf{E}\xi|} + o(1) \right) G_{n|\mathbf{E}\xi|}(x) \quad \text{as } n, x \rightarrow \infty,$$

where $c_\infty \equiv \lim_{k \rightarrow \infty} c_k = \int_{\mathbf{R}} c(u) \pi(du)$.

Remark 2. If the distribution tail of G is regularly varying at infinity and $c_\infty > 0$, then the last theorem implies the asymptotics $\pi_n(x) \sim c_\infty G_\infty(x)/|\mathbf{E}\xi|$ as $n/x \rightarrow \infty$. In particular, in the region $n/x \rightarrow \infty$ the asymptotical behavior of the probability $\pi_n(x)$ coincides with that of the invariant measure tail: $\pi_n(x) \sim \pi(x)$.

Proof of Theorem 3. By virtue of the theorem conditions, for any $\varepsilon > 0$, there exist a random variable ζ_ε with a distribution G_ε and level U_ε such that $\mathbf{E}\zeta_\varepsilon \leq \mathbf{E}\xi + \varepsilon$, $\mathbf{P}\{\zeta_\varepsilon \geq t\} = \mathbf{P}\{\zeta \geq t\}$ for all sufficiently large t , and, for $t \geq U_\varepsilon$,

$$\mathbf{P}\{u + \xi(u) \geq t\} \leq G_\varepsilon(t) \quad \text{if } u < U_\varepsilon, \quad \xi(u) \leq_{\text{st}} \zeta_\varepsilon \quad \text{if } u \geq U_\varepsilon.$$

It follows from Lemma 7 that

$$\pi_n(x) \leq (1 + o(1)) \sum_{k=1}^n c_{k-1} G(x + (n - k) (|\mathbf{E}\xi| - \varepsilon))$$

as $x \rightarrow \infty$ uniformly in $n \geq 1$. Since the function $G(y)$ is long-tailed, it implies, by the arbitrary choice of $\varepsilon > 0$, the following upper estimate:

$$\pi_n(x) \leq (1 + o(1)) \sum_{k=1}^n c_{k-1} G(x + (n - k) |\mathbf{E}\xi|).$$

Since the time parameter n takes only a countable number of values, the corresponding lower estimate follows from Lemma 3 and from the following relation: For any fixed n , $\pi_n(x) \geq (1 + o(1)) G(x) \sum_{k=1}^n c_{k-1}$ as $x \rightarrow \infty$. The latter relation may be verified by induction in the same way as relation (3.12). The proof of the theorem is complete.

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