## 7. Martingales and supermartingales

7.1. Introduction. A variety of inequalities have a significant place in the theory of martingales and supermartingales with the discrete time. The first inequalities were deduced by the founder of the theory of martingales J.L. Doob (see [1]).

In what follows, we denote by $S_{k}, k \geq 1$, a supermartingale defined on a filtered probability space $\left(\Omega, \mathcal{F},\left(\mathcal{F}_{k}\right)_{k \geq 0}, \mathbf{P}\right)$ with $S_{0}=0, \mathcal{F}_{0}=\{\varnothing, \Omega\}$, i.e.

$$
\mathbf{E}\left\{S_{k} \mid \mathcal{F}_{k-1}\right\} \leq S_{k-1}
$$

Put $X_{k}=S_{k}-S_{k-1}, k \geq 0$. Define the random variables $\sigma_{k}^{2}$ by the equalities

$$
\sigma_{k}^{2}=\mathbf{E}\left\{X_{k}^{2} \mid \mathcal{F}_{k-1}\right\} .
$$

Denote

$$
B_{k}^{2}=\sum_{1}^{k} \sigma_{j}^{2}, \quad \bar{S}_{n}=\max _{1 \leq k \leq n} S_{k}, \quad \bar{X}_{n}=\max _{1 \leq k \leq n} X_{k}, \quad A_{t}=\sum_{1}^{n} \mathbf{E}\left|X_{j}\right|^{t} .
$$

Define

$$
Q(x)=\mathbf{P}\left(\bar{X}_{n}>x\right)+\mathbf{P}\left(B_{n}>x\right) .
$$

As in the case of independent summands one can distinguish two types of inequalities: a) moment inequalities, i.e. inequalities for $\mathbf{E} f\left(S_{n}\right)$, where $f$ is a function satisfying some restrictions, b) probability inequalities, i.e. bounds for $\mathbf{P}\left(f\left(S_{n}\right)>x\right)$. The simplest case is that of $f(y)=|y|^{t}, t>0$.
(A) Moment inequalities. We start with so called comparison inequalities for martingales obtained in 1966 . by D.L. Burkholder [2],

$$
\begin{equation*}
c_{t}\left(\sum_{1}^{n} \mathbf{E} X_{k}^{2}\right)^{t / 2}<\mathbf{E}\left|S_{n}\right|^{t}<C_{t}\left(\sum_{1}^{n} \mathbf{E} X_{k}^{2}\right)^{t / 2}, \tag{1}
\end{equation*}
$$

where $c_{t}$ and $C_{t}$ are some constants. Of course, one can write these inequalities in the form

$$
\begin{equation*}
C_{t}^{-1} \mathbf{E}\left|S_{n}\right|^{t}<\mathbf{E}\left(\sum_{1}^{n} X_{k}^{2}\right)^{t / 2}<c_{t}^{-1} \mathbf{E}\left|S_{n}\right|^{t} \tag{2}
\end{equation*}
$$

Inequality (1) extends to martingales the well-known inequalities due to Marcinkiewicz Zygmund [3] for independent random variables. In 1973 Burkholder [4] obtained for martingales the next extension of the Rosenthal inequality [5],

$$
\begin{equation*}
k_{t}\left(D_{t}^{1 / t}+\mathbf{E}^{1 / t} B_{n}^{t}\right)<\mathbf{E}^{1 / t}\left|\widehat{S}_{n}\right|^{t}<K_{t}\left(D_{t}^{1 / t}+\mathbf{E}^{1 / t} B_{n}^{t}\right) \tag{3}
\end{equation*}
$$

where $\widehat{S}_{n}=\max _{1 \leq k \leq n} S_{k}$. The variable $B_{n}$ in (3) is random in contrast to the Rosenthal inequality. Thus, the special problem of estimating the expectation $\mathbf{E} B_{n}^{t}$ arises. If the conditional variances $\sigma_{k}^{2}$ admit the uniform bound

$$
\begin{equation*}
\sigma_{k}^{2}<b_{k}^{2} \tag{4}
\end{equation*}
$$

where $b_{k}^{2}$ is some sequence of constants then

$$
\mathbf{E} B_{n}^{t}<\left(\sum_{1}^{n} b_{j}^{2}\right)^{t / 2}
$$

In Burkholder's paper the constants $k_{t}$ and $K_{t}$ are not estimated. A step forward in this direction was made by P. Hitchenko [6] who proved that

$$
\begin{equation*}
K_{t}<K \frac{t}{\ln t} \tag{5}
\end{equation*}
$$

where $K$ is an absolute constant. In my paper [7] the upper bound of the Burkholder type is deduced for the moments $\mathbf{E}\left\{S_{n}^{t} ; S_{n} \geq 0\right\}$ of supermartingales $S_{n}$ with the constant $K_{t}$ satisfying inequality (5). This bound is discussed comprehensively in Section 7.3. Moreover, the numerical bound for the constant $K$ is obtained in my next paper [8]. A short time later, the latter was sharpened by E.L. Presman [9].

The detailed survey of moment inequalities is contained in [10].
(B) Probability inequalities. As to probability inequalities for martingales the case of bounded martingale differences $X_{k}<L$ or $\left|X_{k}\right|<L$ satisfying in addition condition (4) is studied for the most part. The point is that in this case the generating function of moments $\mathbf{E} e^{h S_{n}}$ admits, in essence, the same bounds as in the case of independent summands $X_{k}$. This allows to get for $\mathbf{P}\left(S_{n}>x\right)$ the bounds of the Hoeffding and Bernstein type. The papers [11,12] were the firsts in this direction. The papers [13-16] are devoted to generalizing the Hoeffding and Bernstein inequalities. The probabilities of large deviations of $S_{n}$ are studied in [17] under condition $\max _{1 \leq k \leq n} \mathbf{E}\left|X_{k}\right|^{t}<\infty$.

In Bentkus's paper [18] the probabilities $\mathbf{E}\left(S_{n}>x\right)$ are estimated in terms $\mathbf{E}\left(Z_{n}>x\right)$, where $Z_{n}$ is a sum of independent identically distributed Bernoulli random variables chosen in a proper way. The bounds obtained are compared with the Hoeffding inequalities.

In my paper [7] the bound of the new type

$$
\mathbf{P}\left(\bar{S}_{n}>x\right)<c(t) x^{-t} \int_{0}^{x} Q\left(\varepsilon_{t} u\right) u^{t-1} d u
$$

was obtained, where $c(t), \varepsilon_{t}$ are constants which are defined below in Section 7.2. In the next sections I comprehensively describe the probability and moment inequalities obtained in my papers $[7,8,19]$.
7.2. Probability inequalities. After appearance in 1971 the Nagaev - Fuk inequalities (see [20]) the problem arose to generalize these inequalities to martingales. The first step to this direct was made by D.Kh. Fuk [21] in 1973 under assumption that for some sequence $\left\{y_{k}\right\}_{k \geq 1}, y_{k}>0$,

$$
\begin{equation*}
\mathbf{E}\left\{X_{k}^{2}\left(y_{k}\right) \mid \mathcal{F}_{k-1}\right\}<d_{k}^{2}, \quad \mathbf{E}\left\{\left(X_{k}^{+}\right)^{t}\left(y_{k}\right) \mid \mathcal{F}_{k-1}\right\}<a_{k} \tag{6}
\end{equation*}
$$

where $d_{k}^{2}$ and $a_{k}$ are constants, $t>2$,

$$
X_{k}(y)=\left\{\begin{array}{ll}
X_{k}, & X_{k} \leq y, \\
0, & X_{k}>y,
\end{array}, \quad X_{k}^{+}(y)=\max \left\{0, X_{k}(y)\right\}\right.
$$

These restrictions can seem too strong. It turned out, however, that they are fulfilled, in particular, for the martingale

$$
\mathbf{E}\left\{\left\|\sum_{j=1}^{n} X_{j}\right\| \mid \mathcal{F}_{k}\right\}
$$

where $X_{j}$ are independent random variables, taking values in a separable Banach space, $\mathcal{F}_{k}$ being $\sigma$-algebra generated by random variables $X_{1}, X_{2}, \ldots, X_{k}$, provided

$$
\mathbf{E}\left\|X_{j}\right\|^{t}<\infty, \quad j \in \overline{1, n}
$$

(see, in this connection, [22-24]).
If the martingale $S_{n}$ does not satisfy Fuk's conditions, one can attain this under some restrictions by means of appropriate transformation $f\left(S_{n}\right)$. As applied to Galton - Watson process, this is made in [25]

In the work [26] one of Fuk's inequalities, which contains normal component, is generalized to Banach space under assumption that

$$
\mathbf{E}\left\|X_{j}\right\|^{3}<\infty, \quad j \in \overline{1, n}
$$

In addition a restriction of the same type as Fuk's one is imposed upon the conditional second moments.

Haesler [27] generalized one of Fuk's inequalities as follows: for any $x, u, v>0$

$$
\begin{equation*}
\mathbf{P}\left(\bar{S}_{n}>x\right)<\sum_{i=1}^{n} \mathbf{P}\left(X_{i}>u\right)+\mathbf{P}\left(B_{n}>v\right)+P_{0}(x, u, v) \tag{7}
\end{equation*}
$$

where

$$
P_{0}(x, u, v)=\exp \left\{\frac{x}{u}\left(1-\ln \left(\frac{x u}{v^{2}}\right)\right)\right\} .
$$

In [28] this result is extended to continuous-time martingales. In [29] $P_{0}$ is replaced by for

$$
\begin{equation*}
P_{1}(x, u, v)=\exp \left\{\frac{x}{u}-\left(\frac{v^{2}}{u^{2}}+\frac{x}{u}\right) \ln \left(\frac{x u}{v^{2}}+1\right)\right\} . \tag{8}
\end{equation*}
$$

The bound (7) is applied in [27] for estimating the rate of convergence in the functional CLT for discrete-time martingales. The probability inequalities in [28] and [29] are used in similar way.

The following inequalities were obtained in my work [7].
Theorem 1. Let $0<\gamma \leq 1$ and $t$ satisfy the condition $t \geq \max \left(e^{6}, e^{4} / \gamma^{2}\right)$. Then for every $y>0$

$$
\begin{equation*}
\mathbf{P}\left(\bar{S}_{n}>y\right)<c(t, \gamma) y^{-t} \int_{0}^{y} Q\left(\varepsilon_{t} u\right) u^{t-1} d u \tag{9}
\end{equation*}
$$

where

$$
\varepsilon_{t}=\frac{\ln t-2 \ln \ln t}{2 t}, c(t, \gamma)=\frac{2 e^{6 \gamma t}}{\gamma}
$$

If $\varepsilon_{t}=\eta / t, \eta>0$, then the inequality

$$
\begin{equation*}
\mathbf{P}\left(\bar{S}_{n}>x\right)<c_{1}(t, \eta) x^{-t} \int_{0}^{x} Q\left(\varepsilon_{t} u\right) u^{t-1} d u \tag{10}
\end{equation*}
$$

holds for every $t>0$, where $c_{1}(t, \eta)=t e^{3 \eta \alpha(\eta)} / \eta \alpha(\eta), \alpha(\eta)=e^{\eta+1}$.
Note that

$$
\frac{1}{x^{t}} \int_{0}^{x} Q\left(\varepsilon_{t} u\right) u^{t-1} d u=\int_{0}^{1} Q\left(\varepsilon_{t} s x\right) s^{t-1} d s
$$

which means that right-hand side of (9) decreases in $x$.
The bound (9) extends the inequality from Theorem 4 of the paper [20] (see also [30, Theorem 1.10]). The inequality (9) is closed in form to the main inequality of the paper [31]. The method of the proof is similar to that which used in the papers [31-33].

Naturally the question arises, how Theorem 1 associates with Haesler's inequality (7). Since the inequalities (7) and (9) strongly differ in form it is not so simply to compare them. It is shown in $[7]$ that (9) is not a corollary of (7). Analogues considerations show that it is impossible to derive Burkholder's inequality (15) (which is a generalization of Rosenthal's one) via that of Haesler. The proof of probability inequalities obtained in [7], was later modified in my paper [8], namely, two statements were selected from the former proof, which makes the presentation more transparent. However they are of interest independently.

Proposition 1. For every $x>0, y>0, \alpha>1$ the following inequality holds,

$$
\begin{equation*}
f(x+\alpha y)<f(x) e^{\alpha(1-\ln \alpha)}+Q(y) \tag{11}
\end{equation*}
$$

where $f(x)=\mathbf{P}\left(\bar{S}_{n} \geq x\right)$.
Proposition 2. If the function $f(x)$ does not increase and for every $x>0, y>0, \alpha>1$ satisfies inequality (11), then for every $\alpha>1 \quad \varepsilon>0$

$$
f(x)<\frac{\omega(\alpha, \varepsilon)}{x^{s(\alpha, \varepsilon)}} \int_{0}^{x} Q(\varepsilon u) u^{s(\alpha, \varepsilon)-1} d u
$$

where $s(\alpha, \varepsilon)=\alpha(\ln \alpha-1) / \ln (1+\alpha \varepsilon), \omega(\alpha, \varepsilon)=(\alpha \varepsilon)^{-1} e^{3 \alpha(\ln \alpha-1)}$.
7.3. Moment inequalities. Moment inequalities are deduced easily from inequality (9) in the same way as for independent random variables (see [25,30-32,34] in this connection).

Indeed, by multiplying both sides of (9) for $t+1$ by $t x^{t-1}$, and integrating with respect to $x$ from 0 to $\infty$, we obtain the following inequalities.

For every $t$ and $\gamma$ such that $t>\max \left(e^{6}, e^{2} / \gamma^{2}\right)-1, \quad 0<\gamma \leq 1$,

$$
\begin{equation*}
\mathbf{E}\left\{\bar{S}_{n}^{t} ; \bar{S}_{n} \geq 0\right\}<c(t+1, \gamma) \varepsilon_{t+1}^{-t}\left(\bar{D}_{t}+\mathbf{E} B_{n}^{t}\right) \tag{12}
\end{equation*}
$$

where

$$
\bar{D}_{t}=\mathbf{E}\left\{\bar{X}_{n}^{t} \bar{X}_{n} \geq 0\right\}
$$

Similarly, if $\varepsilon_{t}=\eta / t$, then for every $t>0$

$$
\begin{equation*}
\mathbf{E}\left\{\bar{S}_{n}^{t} ; \bar{S}_{n} \geq 0\right\}<c_{1}(t+1, \eta) \varepsilon_{t+1}^{-t}\left(\bar{D}_{t}+\mathbf{E} B_{n}^{t}\right) \tag{13}
\end{equation*}
$$

If $\left\{S_{k}\right\}_{k \geq 1}$ being a martingale, the inequalities (12) and (13) remain valid for

$$
\mathbf{E}\left\{\left|\widetilde{S}_{n}\right|^{t} ; \widetilde{S}_{n} \leq 0\right\}
$$

where $\widetilde{S}_{n}=\min _{1 \leq k \leq n} S_{k}$, with replacement of $\bar{D}_{t}$ by

$$
\widetilde{D}_{t}=\mathbf{E}\left\{\left|X_{n}\right|^{t} ; \min _{1 \leq k \leq n} X_{k}<0\right\}
$$

By summing the inequalities for $\mathbf{E}\left\{\bar{S}_{n}^{t} ; \bar{S}_{n} \geq 0\right\}$ and $\mathbf{E}\left\{\left|\widetilde{S}_{n}\right|^{t} ; \widetilde{S}_{n}<0\right\}$, we conclude the following: if $S_{n}$ being a martingale, then for every $t>0$ and $\eta>0$,

$$
\begin{equation*}
\mathbf{E} \widehat{S}_{n}^{t}<c_{t}(\eta)\left(D_{t}+2 \mathbf{E} B_{n}^{t}\right) \tag{14}
\end{equation*}
$$

where $\widehat{S}_{n}=\max _{1 \leq k \leq n}\left|S_{k}\right|, \quad D_{t}=\mathbf{E}\left(\max _{1 \leq k \leq n}\left|X_{k}\right|^{t}\right), \quad c_{t}(\eta)=c_{1}(t+1, \eta)((t+1) / \eta)^{t}$.
If $t>\max \left(e^{6}, e^{4} / \gamma^{2}\right)-1, \quad 0<\gamma \leq 1$, then inequality (14) holds with the constant $c_{t}^{\prime}(\gamma)=c(t+1, \gamma) \varepsilon_{t+1}^{-t}$.

By raising both sides of inequality (14) to the power $1 / t$, we have for $t \geq 1$,

$$
\begin{equation*}
\mathbf{E}^{1 / t} \widehat{S}_{n}^{t}<\widehat{c}_{t}\left(D_{t}^{1 / t}+2^{1 / t} \mathbf{E}^{1 / t} B_{n}^{t}\right) \tag{15}
\end{equation*}
$$

where $\widehat{c}_{t}=c_{t}^{1 / t}(\eta)$. If $t>e^{6} \vee e^{4} / \gamma^{2}$, then one can take $\left(c_{t}^{\prime}(\gamma)\right)^{1 / t}$ as $\widehat{c}_{t}$. As it was said above, the latter inequality was obtained by Burkholder [4] without any explicit expression for the constant $\widehat{c}_{t}$.

Since

$$
\begin{equation*}
\mathbf{P}\left(\bar{X}_{n}>x\right) \leq \sum_{1}^{n} \mathbf{P}\left(X_{j}>x\right) \tag{16}
\end{equation*}
$$

one may replace $\bar{D}_{t}$ in the inequality (12) by

$$
A_{t}^{+}=\sum_{1}^{n} \mathbf{E}\left\{X_{j}^{t} ; X_{j} \geq 0\right\}
$$

Respectively in the inequality (14) one may replace $D_{t}$ by $A_{t}$, making it more close in form to the Rosenthal inequality [5]. Sharp bounds in the Rosenthal inequality for $\mathbf{E} S_{n}^{2 k}$, where $S_{n}$ is the sum independent random variables with zero mean, are given in [35].

The inequalities (9) and (10) allow to estimate $\mathbf{E}\left\{g\left(\bar{S}_{n}\right) ; \bar{S}_{n} \geq 0\right\}$ for more extensive class of functions than power ones. We formulate one of such type possible results: if $a$ differentiable function $g(x)$ with $g(0)=0$ satisfies the condition

$$
\frac{g^{\prime}(x)}{x^{t-2}}<\alpha \frac{g^{\prime}(y)}{y^{t-2}}, t>0, \alpha>0
$$

then for every $\eta>0$

$$
\begin{equation*}
\mathbf{E}\left\{g\left(\bar{S}_{n}\right) ; \bar{S}_{n} \geq 0\right\}<\alpha c_{1}(t, \eta)\left(\mathbf{E}\left\{g\left(\varepsilon_{t}^{-1} \bar{X}_{n}\right) ; \bar{X}_{n} \geq 0\right\}+\mathbf{E} g\left(\varepsilon_{t}^{-1} B_{n}\right)\right) \tag{17}
\end{equation*}
$$

where $\varepsilon_{t}=\frac{\eta}{t}$. Indeed, in view of the second assertion of Theorem 1,

$$
\mathbf{E}\left\{g\left(\bar{S}_{n}\right) ; \bar{S}_{n} \geq 0\right\}<c_{2}(t, \eta) \int_{0}^{\infty}\left(x^{-t} \int_{0}^{x} Q\left(\varepsilon_{t} u\right) u^{t-1} d u\right) g_{1}^{\prime}(u) d x \equiv c_{2}(t, \eta) I_{t}
$$

Changing the order of integration, we find

$$
\begin{aligned}
& I_{t}=\int_{0}^{\infty} u^{t-1} Q\left(\varepsilon_{t} u\right)\left(\int_{u}^{\infty} x^{-t} g^{\prime}(x) d x\right) d u<\alpha \int_{0}^{\infty} Q\left(\varepsilon_{t} u\right) g^{\prime}(u) d u \\
& =-\alpha \int_{0}^{\infty} g(u) d Q\left(\varepsilon_{t} u\right)=\alpha\left(\mathbf{E}\left\{g\left(\varepsilon_{t}^{-1} \bar{X}_{n}\right) ; \bar{X}_{n} \geq 0\right\}+\mathbf{E} g\left(\varepsilon_{t}^{-1} B_{n}\right)\right) .
\end{aligned}
$$

The desired result follows immediately from two last relations. The inequality of type (17) is deduced in [4] under slightly weaker restrictions on the function $g$.

Denote $\beta_{t}(n)=\mathbf{E}\left|X_{n}\right|^{t}$. The following bound is obtained in [19].
Let $\left\{S_{k}\right\}_{0}^{\infty}$ be a martingale. Then for every $t \geq 2$

$$
\begin{equation*}
\mathbf{E}\left|S_{n}\right|^{t} \leq c_{t}\left(\sum_{j=1} n \beta_{t}^{2 / t}(j)\right)^{t / 2}, \tag{18}
\end{equation*}
$$

where

$$
\begin{equation*}
c_{t}=\left(\frac{t(t-1)}{2}\right)^{t / 2} . \tag{19}
\end{equation*}
$$

Note that the bound in (18) is achieved under $t=2$. For independent random variables the inequality (18) was obtained by Whittle [36] with the constant

$$
c_{t}^{\prime}=\frac{2^{3 t / 2}}{\sqrt{\pi}} \Gamma\left(\frac{t+1}{2}\right) .
$$

Note that

$$
c_{4}^{\prime} \approx 36.11 \Gamma\left(\frac{5}{2}\right)>c_{4}=36
$$

but

$$
c_{5}^{\prime} \approx 204.31<c_{5}=10^{2.5} \approx 316.22
$$

The following bound extends to martingales the inequality of Dharmadhikari and Jogdeo [37] (see, also, [38, P. 98]).

Let $\beta_{t}(j)=\beta_{t}, 1 \leq j \leq n$. Then for $t>2$

$$
\mathbf{E}\left|S_{n}\right|^{t}<c_{t} n^{t / 2} \beta_{t}
$$

$c_{t}$ being from (19).
Using the well-known inequality

$$
\mathbf{E} \widehat{S}_{n}^{t}<\left(\frac{t}{t-1}\right)^{t} \mathbf{E}\left|S_{n}\right|^{t}
$$

(see, for example, [39, P. 526, Theorem 2] ), we get the following statement: for every $t \geq 2$

$$
\begin{equation*}
\mathbf{E} \widehat{S}_{n}^{t}<c_{t}\left(\sum_{j=1}^{n} \beta_{t}(j)\right)^{t / 2} \tag{20}
\end{equation*}
$$

where $c_{t}=\left(\frac{t^{3}}{t-1}\right)^{t / 2}$.
If $1 \leq t<2$, then the following bound for martingale $S_{n}$ holds,

$$
\mathbf{E}\left|S_{n}\right|^{t}<c_{t} \sum_{j=1}^{n} \beta_{t}(j)
$$

(see [40]).
Note that it is not difficult to get inequalities of type (20) for $\mathbf{E}\left|\widehat{S}_{n}\right|^{2 k}$ from (15). Indeed, using the inequality

$$
\mathbf{E} \prod_{l=1}^{m} \sigma_{l}^{2 i_{l}} \leq \prod_{l=1}^{m} \beta_{2 k}^{i_{l} / 2 k}(l)
$$

we have

$$
\mathbf{E} B_{n}^{2 k}<\left(\sum_{j=1}^{n} \beta_{2 k}^{1 / k}(j)\right)^{k}
$$

Hence,

$$
\mathbf{E}\left|\widehat{S}_{n}\right|^{2 k}<c_{2 k}\left(A_{2 k}+2\left(\sum_{j=1}^{n} \beta_{2 k}^{1 / k}(j)\right)^{k}\right)
$$

where $c_{2 k}=\min _{\eta} c_{2 k}(\eta)$. Making this estimate crude, we can write

$$
\mathbf{E} \widehat{S}_{n}^{2 k}<3 c_{2 k}\left(\sum_{j=1}^{n} \beta_{2 k}^{1 / k}(j)\right)^{k}
$$

7.4. Numerical estimates. P. Hitchenko [6] had shown that

$$
\begin{equation*}
\mathbf{E} \widehat{S}_{n}^{t}<K \frac{t}{\ln t}\left(D_{t}^{1 / t}+\mathbf{E}^{1 / t} B_{n}^{t}\right) \tag{21}
\end{equation*}
$$

where $K$ is an absolute constant (see also [41]).
In connection with the inequality (21) the bound of the quantity

$$
K_{t}=\frac{\mathbf{E}^{1 / t}\left\{\bar{S}_{n}^{t} ; S_{n} \geq 0\right\}}{\bar{D}_{t}^{1 / t}+\mathbf{E}^{1 / t} B_{n}^{t}} \frac{\ln t}{t}
$$

is of interest.
Putting $\widehat{c}_{t}=\left(c_{t}^{\prime}(\gamma)\right)^{1 / t}$ in (15), we arrive to the bound

$$
\limsup _{t \rightarrow \infty} K_{t} \leq 2 e^{6 \gamma}
$$

Since $\gamma$ can be made arbitrarily small, we have the right to state that

$$
\begin{equation*}
\limsup _{t \rightarrow \infty} K_{t} \leq 2 \tag{22}
\end{equation*}
$$

It was shown in [42] that

$$
\lim _{t \rightarrow \infty} \widehat{c}_{t} \frac{\ln t}{t}=\frac{1}{e}
$$

if $X_{k}$ are independent symmetrically distributed random variables.
Estimates for $K_{t}$ under $t>2$ are given in my work [8].
Let us introduce the definitions,

$$
\begin{aligned}
g_{0}(t) & =\left(1+\frac{1}{t}\right) \ln (t+1)-\ln t+\ln \ln t \\
g_{1}(t) & =g_{0}(t)+\frac{2.76}{t}+1.39 \\
g_{2}(t) & =g_{0}(t)+\frac{3.694}{t}+1.064, \\
g_{3}(t) & =g_{0}(t)+\frac{1.74}{\ln t-1.1}-\left(1+\frac{1}{t}\right) \ln (\ln t-2.86)+0.57\left(1+\frac{1}{t}\right)+ \\
& +\frac{0.57}{t}(2.86-\ln t) \quad \text { if } \quad \ln t>2.86
\end{aligned}
$$

$$
g_{3}(t)=\infty \quad \text { if } \quad \ln t \leq 2.86
$$

The following result was obtained in [8]: for every $t>2$

$$
\begin{equation*}
K_{t}<g(t):=\min _{1 \leq j \leq 3} e^{g_{j}(t)} \tag{23}
\end{equation*}
$$

Analysis of behavior of the functions $g_{i}(t), i=\overline{1,3}$, leads to the following deduction (see also Fig. 1 and 2). First of all, $g_{1}(t)<g_{2}(t)$ for $2<t<t_{0}=2.865 \ldots, g_{1}(t)>g_{2}(t)$ for $t>t_{0}$.


Fig. 1. $g_{1}(t)<g_{2}(t), 2<t<t_{0}$;

$$
g_{1}(t)>g_{2}(t), t>t_{0}
$$



Fig. 2. $g_{3}(t)<g_{2}(t), t>t_{2}$

As to the function $g_{3}$, it decreases, where $g_{3}(t)>g_{2}(t)$ for $e^{2.86} \equiv 17.46 \ldots<t<t_{2}=$ $49.936 \ldots, g_{3}(t)<g_{2}(t)$ for $t>t_{2}, t_{2}$ being the root of the equation $g_{2}(t)=g_{3}(t)$. Thus,

$$
K_{t}<\left\{\begin{array}{lc}
e^{g_{1}(t)}, & 2<t<3, \\
e^{g_{2}(t)}, & 3 \leq t<t_{2}, \\
e^{g_{3}(t)}, & t>t_{2}
\end{array}\right.
$$

In particular, the next result follows from the aforesaid,

$$
\begin{equation*}
\sup _{2<t<3} K_{t} \leq e^{g_{1}(2)} \approx 29, \sup _{3 \leq t<4} K_{t} \leq e^{g_{2}(3)} \approx 23.1, \sup _{t \geq 4} K_{t} \leq e^{g_{2}(4)} \approx 18.9 \tag{24}
\end{equation*}
$$

It is easily seen that $g_{3}(t) \rightarrow 0.57$ for $t \rightarrow \infty$. Consequently,

$$
\limsup _{t \rightarrow \infty} K_{t} \leq e^{0.57}<1.77
$$

Similar bounds for independent $X_{i}$ were obtained in [43] as well as in [44].
Recently Presman [9] proved that

$$
\limsup _{t \rightarrow \infty} K_{t} \leq 1, \quad K_{t} \leq 9.46
$$

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