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 , $k \ge 1$, a supermartingale defined on a filtered probability space $(\Omega, \mathcal{F}, (\mathcal{F}_k)_{k\ge 0}, \mathbf{P})$ with $S_0 = 0$, $\mathcal{F}_0 = \{\emptyset, \Omega\}$, i.e.

$$\mathbf{E}\{S_k \,|\, \mathcal{F}_{k-1}\} \le S_{k-1}.$$

Put $X_k = S_k - S_{k-1}, k \ge 0$. Define the random variables σ_k^2 by the equalities

$$\sigma_k^2 = \mathbf{E}\{X_k^2 \,|\, \mathcal{F}_{k-1}\}.$$

Denote

$$B_k^2 = \sum_{1}^{k} \sigma_j^2, \quad \overline{S}_n = \max_{1 \le k \le n} S_k, \quad \overline{X}_n = \max_{1 \le k \le n} X_k, \quad A_t = \sum_{1}^{n} \mathbf{E} |X_j|^t.$$

Define

$$Q(x) = \mathbf{P}(\overline{X}_n > x) + \mathbf{P}(B_n > x).$$

As in the case of independent summands one can distinguish two types of inequalities: **a)** moment inequalities, i.e. inequalities for $\mathbf{E}f(S_n)$, where f is a function satisfying some restrictions, **b)** probability inequalities, i.e. bounds for $\mathbf{P}(f(S_n) > x)$. 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],

$$c_t \left(\sum_{1}^{n} \mathbf{E} X_k^2\right)^{t/2} < \mathbf{E} |S_n|^t < C_t \left(\sum_{1}^{n} \mathbf{E} X_k^2\right)^{t/2},\tag{1}$$

where c_t and C_t are some constants. Of course, one can write these inequalities in the form

$$C_t^{-1} \mathbf{E} |S_n|^t < \mathbf{E} \Big(\sum_{1}^n X_k^2\Big)^{t/2} < c_t^{-1} \mathbf{E} |S_n|^t.$$
 (2)

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],

$$k_t \left(D_t^{1/t} + \mathbf{E}^{1/t} B_n^t \right) < \mathbf{E}^{1/t} \big| \widehat{S}_n \big|^t < K_t \left(D_t^{1/t} + \mathbf{E}^{1/t} B_n^t \right), \tag{3}$$

where $\widehat{S}_n = \max_{1 \le k \le n} S_k$. 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 σ_k^2 admit the uniform bound

$$\sigma_k^2 < b_k^2,\tag{4}$$

where b_k^2 is some sequence of constants then

$$\mathbf{E}B_n^t < \Big(\sum_{1}^n b_j^2\Big)^{t/2}.$$

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

$$K_t < K \frac{t}{\ln t},\tag{5}$$

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 $|X_k| < L$ satisfying in addition condition (4) is studied for the most part. The point is that in this case the generating function of moments $\mathbf{E}e^{hS_n}$ admits, in essence, the same bounds as in the case of independent summands X_k . This allows to get for $\mathbf{P}(S_n > x)$ 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 $\max_{1 \le k \le n} \mathbf{E}|X_k|^t < \infty$.

In Bentkus's paper [18] the probabilities $\mathbf{E}(S_n > x)$ are estimated in terms $\mathbf{E}(Z_n > x)$, 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

$$\mathbf{P}(\overline{S}_n > x) < c(t)x^{-t} \int_0^x Q(\varepsilon_t u) u^{t-1} \, du$$

was obtained, where c(t), ε_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 $\{y_k\}_{k\geq 1}, y_k > 0$,

$$\mathbf{E}\{X_{k}^{2}(y_{k}) | \mathcal{F}_{k-1}\} < d_{k}^{2}, \qquad \mathbf{E}\{(X_{k}^{+})^{t}(y_{k}) | \mathcal{F}_{k-1}\} < a_{k},$$
(6)

where d_k^2 and a_k are constants, t > 2,

$$X_k(y) = \begin{cases} X_k, & X_k \le y, \\ 0, & X_k > y, \end{cases}, \quad X_k^+(y) = \max\{0, X_k(y)\}$$

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

$$\mathbf{E}\bigg\{\bigg\|\sum_{j=1}^{n}X_{j}\bigg\|\bigg|\mathcal{F}_{k}\bigg\},\$$

where X_j are independent random variables, taking values in a separable Banach space, \mathcal{F}_k being σ -algebra generated by random variables X_1, X_2, \ldots, X_k , provided

$$\mathbf{E} \|X_j\|^t < \infty, \qquad j \in \overline{1, n}$$

(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 $f(S_n)$. 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

$$\mathbf{E} \|X_j\|^3 < \infty, \qquad j \in \overline{1, n}.$$

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

$$\mathbf{P}(\overline{S}_n > x) < \sum_{i=1}^n \mathbf{P}(X_i > u) + \mathbf{P}(B_n > v) + P_0(x, u, v),$$
(7)

where

$$P_0(x, u, v) = \exp\left\{\frac{x}{u}\left(1 - \ln\left(\frac{xu}{v^2}\right)\right)\right\}.$$

In [28] this result is extended to continuous-time martingales. In [29] P_0 is replaced by for

$$P_1(x, u, v) = \exp\left\{\frac{x}{u} - \left(\frac{v^2}{u^2} + \frac{x}{u}\right) \ln\left(\frac{xu}{v^2} + 1\right)\right\}.$$
(8)

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 $t \geq \max(e^6, e^4/\gamma^2)$. Then for every y > 0

$$\mathbf{P}(\overline{S}_n > y) < c(t, \gamma) y^{-t} \int_0^y Q(\varepsilon_t u) u^{t-1} du,$$
(9)

where

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

If $\varepsilon_t = \eta/t$, $\eta > 0$, then the inequality

$$\mathbf{P}(\overline{S}_n > x) < c_1(t,\eta) \, x^{-t} \int_0^x Q(\varepsilon_t u) \, u^{t-1} \, du \tag{10}$$

holds for every t > 0, where $c_1(t,\eta) = te^{3\eta\alpha(\eta)}/\eta\alpha(\eta)$, $\alpha(\eta) = e^{\eta+1}$.

Note that

$$\frac{1}{x^t} \int_0^x Q(\varepsilon_t u) \, u^{t-1} \, du = \int_0^1 Q(\varepsilon_t s x) \, s^{t-1} \, ds$$

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,

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

where $f(x) = \mathbf{P}(\overline{S}_n \ge x)$.

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 $\alpha > 1$ $\varepsilon > 0$

$$f(x) < \frac{\omega(\alpha,\varepsilon)}{x^{s(\alpha,\varepsilon)}} \int_0^x Q(\varepsilon u) \, u^{s(\alpha,\varepsilon)-1} \, du,$$

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 tx^{t-1} , and integrating with respect to x from 0 to ∞ , we obtain the following inequalities.

For every t and γ such that $t > \max(e^6, e^2/\gamma^2) - 1, \ 0 < \gamma \le 1,$

$$\mathbf{E}\{\overline{S}_{n}^{t}; \ \overline{S}_{n} \ge 0\} < c(t+1,\gamma) \varepsilon_{t+1}^{-t}(\overline{D}_{t} + \mathbf{E}B_{n}^{t}),$$
(12)

where

$$\overline{D}_t = \mathbf{E}\{\overline{X}_n^t \ \overline{X}_n \ge 0\}.$$

Similarly, if $\varepsilon_t = \eta/t$, then for every t > 0

$$\mathbf{E}\{\overline{S}_{n}^{t}; \ \overline{S}_{n} \ge 0\} < c_{1}(t+1,\eta) \varepsilon_{t+1}^{-t}(\overline{D}_{t} + \mathbf{E}B_{n}^{t}).$$
(13)

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

 $\mathbf{E}\{|\widetilde{S}_n|^t; \ \widetilde{S}_n \le 0\},\$

where $\widetilde{S}_n = \min_{1 \le k \le n} S_k$, with replacement of \overline{D}_t by

$$\widetilde{D}_t = \mathbf{E}\big\{|X_n|^t; \min_{1 \le k \le n} X_k < 0\big\}.$$

By summing the inequalities for $\mathbf{E}\{\overline{S}_n^t; \overline{S}_n \ge 0\}$ and $\mathbf{E}\{|\widetilde{S}_n|^t; \widetilde{S}_n < 0\}$, we conclude the following: if S_n being a martingale, then for every t > 0 and $\eta > 0$,

$$\mathbf{E}\widehat{S}_{n}^{t} < c_{t}(\eta)(D_{t} + 2\mathbf{E}B_{n}^{t}), \tag{14}$$

where $\widehat{S}_n = \max_{1 \le k \le n} |S_k|$, $D_t = \mathbf{E}(\max_{1 \le k \le n} |X_k|^t)$, $c_t(\eta) = c_1(t+1, \eta)((t+1)/\eta)^t$. If $t > \max(e^6, e^4/\gamma^2) - 1$, $0 < \gamma \le 1$, then inequality (14) holds with the constant $c'_t(\gamma) = c(t+1, \gamma) \varepsilon_{t+1}^{-t}$.

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

$$\mathbf{E}^{1/t}\widehat{S}_{n}^{t} < \widehat{c}_{t}(D_{t}^{1/t} + 2^{1/t}\mathbf{E}^{1/t}B_{n}^{t}),$$
(15)

where $\hat{c}_t = c_t^{1/t}(\eta)$. If $t > e^6 \vee e^4/\gamma^2$, then one can take $(c_t'(\gamma))^{1/t}$ as \hat{c}_t . As it was said above, the latter inequality was obtained by Burkholder [4] without any explicit expression for the constant \hat{c}_t .

Since

$$\mathbf{P}(\overline{X}_n > x) \le \sum_{1}^{n} \mathbf{P}(X_j > x), \tag{16}$$

one may replace \overline{D}_t in the inequality (12) by

$$A_t^+ = \sum_{1}^{n} \mathbf{E}\{X_j^t; X_j \ge 0\}$$

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 $\mathbf{E}S_n^{2k}$, where S_n is the sum independent random variables with zero mean, are given in [35].

The inequalities (9) and (10) allow to estimate $\mathbf{E}\{g(\overline{S}_n); \overline{S}_n \geq 0\}$ 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

$$\frac{g'(x)}{x^{t-2}} < \alpha \frac{g'(y)}{y^{t-2}}, \ t > 0, \ \alpha > 0,$$

then for every $\eta > 0$

$$\mathbf{E}\{g(\overline{S}_n); \overline{S}_n \ge 0\} < \alpha c_1(t, \eta) \Big(\mathbf{E}\{g(\varepsilon_t^{-1} \overline{X}_n); \overline{X}_n \ge 0\} + \mathbf{E}g(\varepsilon_t^{-1} B_n) \Big),$$
(17)

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

$$\mathbf{E}\{g(\overline{S}_n); \overline{S}_n \ge 0\} < c_2(t,\eta) \int_0^\infty \left(x^{-t} \int_0^x Q(\varepsilon_t u) u^{t-1} du\right) g_1'(u) dx \equiv c_2(t,\eta) I_t.$$

Changing the order of integration, we find

$$I_t = \int_0^\infty u^{t-1} Q(\varepsilon_t u) \left(\int_u^\infty x^{-t} g'(x) \, dx \right) du < \alpha \int_0^\infty Q(\varepsilon_t u) g'(u) \, du$$
$$= -\alpha \int_0^\infty g(u) \, dQ(\varepsilon_t u) = \alpha \left(\mathbf{E} \{ g(\varepsilon_t^{-1} \overline{X}_n); \overline{X}_n \ge 0 \} + \mathbf{E} g(\varepsilon_t^{-1} B_n) \right).$$

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 $\beta_t(n) = \mathbf{E} |X_n|^t$. The following bound is obtained in [19].

Let $\{S_k\}_0^\infty$ be a martingale. Then for every $t \ge 2$

$$\mathbf{E}|S_n|^t \le c_t \left(\sum_{j=1}^{t} n\beta_t^{2/t}(j)\right)^{t/2},\tag{18}$$

where

$$c_t = \left(\frac{t(t-1)}{2}\right)^{t/2}.$$
(19)

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

$$c_t' = \frac{2^{3t/2}}{\sqrt{\pi}} \Gamma\left(\frac{t+1}{2}\right).$$

Note that

$$c_4' \approx 36.11 \, \Gamma\left(\frac{5}{2}\right) > c_4 = 36$$

but

$$c_5' \approx 204.31 < c_5 = 10^{2.5} \approx 316.22$$

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

Let $\beta_t(j) = \beta_t$, $1 \le j \le n$. Then for t > 2

$$\mathbf{E}|S_n|^t < c_t n^{t/2} \beta_t,$$

 c_t being from (19).

Using the well-known inequality

$$\mathbf{E}\widehat{S}_n^t < \left(\frac{t}{t-1}\right)^t \mathbf{E}|S_n|^t$$

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

$$\mathbf{E}\widehat{S}_{n}^{t} < c_{t} \Big(\sum_{j=1}^{n} \beta_{t}(j)\Big)^{t/2}, \tag{20}$$

where $c_t = \left(\frac{t^3}{t-1}\right)^{t/2}$. If $1 \le t \le 2$, then

If $1 \le t < 2$, then the following bound for martingale S_n holds,

$$\mathbf{E}|S_n|^t < c_t \sum_{j=1}^n \beta_t(j)$$

(see [40]).

Note that it is not difficult to get inequalities of type (20) for $\mathbf{E}|\hat{S}_n|^{2k}$ from (15). Indeed, using the inequality

$$\mathbf{E}\prod_{l=1}^m \sigma_l^{2i_l} \le \prod_{l=1}^m \beta_{2k}^{i_l/2k}(l),$$

we have

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

Hence,

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

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

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

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

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

where K is an absolute constant (see also [41]).

In connection with the inequality (21) the bound of the quantity

$$K_t = \frac{\mathbf{E}^{1/t} \{ \overline{S}_n^t; \ S_n \ge 0 \}}{\overline{D}_t^{1/t} + \mathbf{E}^{1/t} B_n^t} \frac{\ln t}{t}$$

is of interest.

Putting $\hat{c}_t = (c'_t(\gamma))^{1/t}$ in (15), we arrive to the bound

$$\limsup_{t \to \infty} K_t \le 2e^{6\gamma}.$$

Since γ can be made arbitrarily small, we have the right to state that

$$\limsup_{t \to \infty} K_t \le 2. \tag{22}$$

It was shown in [42] that

$$\lim_{t \to \infty} \widehat{c}_t \frac{\ln t}{t} = \frac{1}{e}$$

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,

$$g_{0}(t) = \left(1 + \frac{1}{t}\right) \ln(t+1) - \ln t + \ln \ln t,$$

$$g_{1}(t) = g_{0}(t) + \frac{2.76}{t} + 1.39,$$

$$g_{2}(t) = g_{0}(t) + \frac{3.694}{t} + 1.064,$$

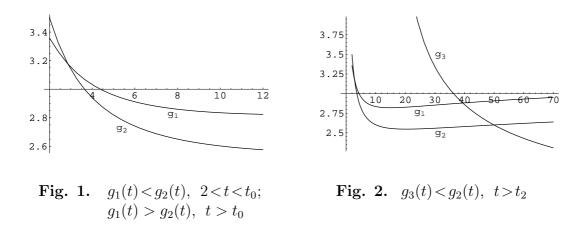
$$g_{3}(t) = g_{0}(t) + \frac{1.74}{\ln t - 1.1} - \left(1 + \frac{1}{t}\right) \ln(\ln t - 2.86) + 0.57\left(1 + \frac{1}{t}\right) + \frac{0.57}{t}\left(2.86 - \ln t\right) \quad \text{if} \quad \ln t > 2.86,$$

 $g_3(t) = \infty \quad \text{if} \quad \ln t \le 2.86.$

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

$$K_t < g(t) := \min_{1 \le j \le 3} e^{g_j(t)}.$$
(23)

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 $2 < t < t_0 = 2.865...$, $g_1(t) > g_2(t)$ for $t > t_0$.



As to the function g_3 , it decreases, where $g_3(t) > g_2(t)$ for $e^{2.86} \equiv 17.46 \dots < t < t_2 = 49.936 \dots$, $g_3(t) < g_2(t)$ for $t > t_2$, t_2 being the root of the equation $g_2(t) = g_3(t)$. Thus,

$$K_t < \begin{cases} e^{g_1(t)}, & 2 < t < 3, \\ e^{g_2(t)}, & 3 \le t < t_2, \\ e^{g_3(t)}, & t > t_2. \end{cases}$$

In particular, the next result follows from the aforesaid,

$$\sup_{2 < t < 3} K_t \le e^{g_1(2)} \approx 29, \ \sup_{3 \le t < 4} K_t \le e^{g_2(3)} \approx 23.1, \ \sup_{t \ge 4} K_t \le e^{g_2(4)} \approx 18.9.$$
(24)

It is easily seen that $g_3(t) \to 0.57$ for $t \to \infty$. Consequently,

$$\limsup_{t \to \infty} K_t \le e^{0.57} < 1.77.$$

Similar bounds for independent X_i were obtained in [43] as well as in [44].

Recently Presman [9] proved that

$$\limsup_{t \to \infty} K_t \le 1, \quad K_t \le 9.46.$$

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