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# Comparison inequalities for suprema of bounded empirical processes

#### Antoine Marchina\*

#### Abstract

In this Note we provide comparison moment inequalities for suprema of bounded empirical processes. Our methods are only based on a decomposition in martingale and on comparison results concerning martingales proved by Bentkus and Pinelis.

**Keywords:** concentration inequalities; generalized moments; empirical processes; martingale method.

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#### **1** Introduction

Let  $X_1, \ldots, X_n$  be a sequence of independent random variables valued in some measurable space  $(\mathcal{X}, \mathcal{F})$  with common distribution P. Let  $P_n$  denote for every integer n the empirical probability measure  $P_n := n^{-1}(\delta_{X_1} + \ldots + \delta_{X_n})$ . Let  $\mathscr{F}$  be a countable class of measurable functions  $f : \mathcal{X} \to \mathbb{R}$  such that P(f) = 0 for all  $f \in \mathscr{F}$ . In this Note, we are concerned with concentration properties around the mean of the random variable

$$Z := \sup\{nP_n(f) : f \in \mathscr{F}\},\tag{1.1}$$

when  $\mathscr{F}$  satisfies a two-sided or a one-sided (from above) boundedness condition. Our approach is based on a decomposition of  $Z - \mathbb{E}[Z]$  into a sum of martingale increments together with comparison inequalities for martingales with (two-sided or one-sided) bounded increments proved by Bentkus [1] and Pinelis [7]. Before going further, let us introduce some notations.

**Definition 1.1.** (i) Let  $\alpha, \beta$  be two reals such that  $\alpha < \beta$ . We say that a random variable  $\theta$  follows a Bernoulli distribution if it assumes exactly two values and we write  $\theta \sim \mathcal{B}_m(\alpha, \beta)$  if

$$\mathbb{P}(\theta = \beta) = 1 - \mathbb{P}(\theta = \alpha) \in [0, 1[, \text{ and } \mathbb{E}[\theta] = m.$$
(1.2)

Notice that

$$Var(\theta) = (m - \alpha)(\beta - m).$$
(1.3)

(ii) For any  $a \ge 0$ ,  $\Gamma_a$  stands for any centered Gaussian random variable with variance equals to a.

(iii) For any a > 0,  $\Pi_a$  stands for any Poisson random variable with parameter a. We also denote by  $\tilde{\Pi}_a$  the centered Poisson random variable  $\tilde{\Pi}_a := \Pi_a - a$ .

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#### Comparison inequalities for suprema of bounded empirical processes

Let us introduce the class of convex functions in which the comparison inequalities, stated in this Note, are valid.

**Definition 1.2.** Let  $k \in \mathbb{N}^*$ . As usual, we denote by  $\mathcal{C}^k$  the space of k-times continuously differentiable functions from  $\mathbb{R}$  to  $\mathbb{R}$ . We define the following class of functions:

$$\mathcal{G}^{k} := \{ \varphi \in \mathcal{C}^{k-1} : \varphi, \, \varphi', \dots, \varphi^{(k-1)} \text{ are convex} \}.$$
(1.4)

Furthermore, we use the classical notations  $x_+ := \max(0, x)$  and  $x_+^{\alpha} = (x_+)^{\alpha}$ . We now recall the two following comparison results, proved by Bentkus [1] and Pinelis [7], which are the main tools in our proofs.

**Proposition 1.3** (Lemmas 4.4 and 4.5 in [1] and Theorem 2.1 and Remark 2.3 in [7]). Let  $s_1^2, \ldots, s_n^2, \beta_1, \ldots, \beta_n$  be positive reals. Let  $M_n := \sum_{k=1}^n X_k$  be a martingale with respect to a nondecreasing filtration ( $\mathcal{F}_k$ ) such that  $M_0 = 0$ ,

$$X_k \le 1$$
, and  $\mathbb{E}[X_k^2 \mid \mathcal{F}_{k-1}] \le s_k^2$  a.s. (1.5)

(i) Let  $s^2 := n^{-1}(s_1^2 + \ldots + s_n^2)$  and  $S_n := \theta_1 + \ldots + \theta_n$  be a sum of n independent copies of a random variable  $\theta$  with distribution  $\mathcal{B}_0(-s^2, 1)$  (defined by (1.2)). Then for any  $\varphi \in \mathcal{G}^2$ ,

$$\mathbb{E}[\varphi(M_n)] \le \mathbb{E}[\varphi(S_n)]. \tag{a}$$

(ii) Additionally to (1.5), assuming that

$$\mathbb{E}_{k-1}[X_{k+}^3] \le \beta_k \text{ a.s., and } \beta := \frac{1}{n} \sum_{k=1}^n \beta_k \le s^2,$$
(1.6)

we have for any  $\varphi \in \mathcal{G}^3$ ,

$$\mathbb{E}[\varphi(M_n)] \le \mathbb{E}[\varphi(\Gamma_{n(s^2 - \beta)} + \tilde{\Pi}_{n\beta})], \qquad (b)$$

where  $\Gamma_{n(s^2-\beta)}$  and  $\Pi_{n\beta}$  are independent and respectively defined by Definition 1.1 (ii) and (iii).

**Remark 1.4.** In fact, the results in the original papers are stated in the following slightly smaller class of functions:

$$\begin{aligned} \mathcal{H}^{\alpha}_{+} &:= \big\{ \varphi : \varphi(u) = \int_{-\infty}^{\infty} (u-t)^{\alpha}_{+} \, \mu(dt) \text{ for some Borel measure } \mu \geq 0 \text{ on } \mathbb{R} \\ & \text{ and all } u \in \mathbb{R} \big\}, \end{aligned}$$

for  $\alpha \in \{2,3\}$ . The extensions to  $\mathcal{G}^{\alpha}$  follow from a result of Pinelis [8, Corollary 5.8] (see also [6, Section 2]).

**Remark 1.5.** From moment comparison inequalities in  $\mathcal{H}^{\alpha}_{+}$ , such as in the above Proposition, one can derive tail comparison inequalities. We refer the reader to [4, 5, 6] for the statements of these results and for some more details.

Finally, we use the notations:

$$E_k := \mathbb{E} \sup_{f \in \mathscr{F}} P_k(f) \text{ for any } k = 1, \dots, n \text{ and } \bar{E} := n^{-1}(E_1 + \dots + E_n).$$
(1.7)

#### 2 Results

#### 2.1 Two-sided boundedness condition

Here, by two-sided boundedness condition, we mean that  $\mathscr{F}$  is a countable class of measurable functions with values in [-a, 1] for some positive real a. Let  $\psi$  be the function defined on [0, 1] by

$$\psi(x) = x(1-x)$$
 if  $x \in [0, 1/2]$  and  $\psi(x) = 1/4$  if  $x \in [1/2, 1]$ . (2.1)

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**Theorem 2.1.** Let  $\mathscr{F}$  be a countable class of measurable functions from  $\mathcal{X}$  into [-a, 1] such that P(f) = 0 for all  $f \in \mathscr{F}$ . Let Z be defined by (1.1). (i) Case a > 1.

Let  $\theta$  be a Bernoulli random variable with distribution  $\mathcal{B}_0(-a,1)$  (defined by (1.2)). Let  $\theta_1, \ldots, \theta_n$  be n independent copies of  $\theta$  and let  $S_n := \theta_1 + \ldots + \theta_n$ . Then for any function  $\varphi \in \mathcal{G}^2$ ,

$$\mathbb{E}[\varphi(Z - \mathbb{E}Z)] \le \mathbb{E}[\varphi(S_n)]. \tag{a}$$

(*ii*) Case a < 1.

Let  $\vartheta$  be a Bernoulli random variable with distribution

$$\mathcal{B}_0\Big(-(a+1)^2\,\psi\Big(\frac{a+\bar{E}}{a+1}\Big),\,1\Big)$$

(defined by (1.2)), where  $\psi$  is defined by (2.1), and  $\overline{E}$  is defined by (1.7). Let  $\vartheta_1, \ldots, \vartheta_n$  be *n* independent copies of  $\vartheta$  and let  $T_n := \vartheta_1 + \ldots + \vartheta_n$ . Then for any function  $\varphi \in \mathcal{G}^2$ ,

$$\mathbb{E}[\varphi(Z - \mathbb{E}Z)] \le \mathbb{E}[\varphi(T_n)]. \tag{b}$$

**Remark 2.2.** If the class  $\mathscr{F}$  satisfies the uniform law of large numbers, that is the random variable  $\sup_{f \in \mathscr{F}} |P_n(f)|$  converges to 0 in probability, then  $E_n$  decreases to 0 (see, for instance, Section 2.4 of van der Vaart and Wellner [10]) and so  $\overline{E}$  also decreases to 0. This ensures that  $n \mapsto (a+1)^2 \psi((a+\overline{E})/(a+1))$  (which is also the variance of  $\vartheta$ ) is nonincreasing and tends to a as n tends to infinity.

**Example 2.3** (Set-indexed empirical processes). Let  $\mathscr{S}$  be a countable class of measurable sets of  $\mathcal{X}$ . We consider the class of functions

$$\mathscr{F} = \left\{ \frac{\mathbb{1}_S - P(S)}{1 - P(S)} : S \in \mathscr{S} \right\}.$$
(2.2)

Let  $p := \sup\{P(S) : S \in \mathscr{S}\}$  and we assume that p < 1. Since  $x \mapsto x/(1-x)$  is increasing on [0,1[, we can apply Theorem 2.1 with a = p/(1-p). Hence with the notations of Theorem 2.1:

(i) If p > 1/2, then for any  $\varphi \in \mathcal{G}^2$ ,  $\mathbb{E}[\varphi(Z - \mathbb{E}[Z])] \leq \mathbb{E}[\varphi(S_n)]$ , and  $\theta$  has the distribution  $\mathcal{B}_0(-p/(1-p), 1)$ .

(ii) If p < 1/2, then for any  $\varphi \in \mathcal{G}^2$ ,  $\mathbb{E}[\varphi(Z - \mathbb{E}[Z])] \leq \mathbb{E}[\varphi(T_n)]$ , and  $\vartheta$  has the distribution  $\mathcal{B}_0(-(1-p)^{-2}\psi(\bar{E}+p(1-\bar{E})), 1)$ .

Theorem 3.1 of Rio [9], when applied to Z (see also his Theorem 4.2 (a)), provides a Bennett-type inequality for classes of sets with small measures under P. Precisely the condition is  $E_n + p(1 - E_n) \le 1/2$ . Hence since  $\mathcal{G}^2$  contains all increasing exponential functions  $x \mapsto e^{tx}$ , t > 0, the Case (*ii*) above completes Rio's result in this situation.

#### 2.2 One-sided boundedness condition

Here, by one-sided boundedness condition, we mean that  $\mathscr{F}$  is a countable class of measurable functions with values in  $]-\infty, 1]$ . Let  $\rho$  be the function defined on [0, 1] by

$$\rho(x) = x(1-x)^2 \text{ if } x \in [0, 1/3] \text{ and } \rho(x) = 4/27 \text{ if } x \in [1/3, 1].$$
(2.3)

**Theorem 2.4.** Let  $\mathscr{F}$  be a countable class of measurable functions from  $\mathcal{X}$  into  $]-\infty, 1]$  such that P(f) = 0 for any  $f \in \mathscr{F}$ . Let Z be defined by (1.1). Define also

$$\sigma^{2} := \sup\{P(f^{2}) : f \in \mathscr{F}\} \text{ and } m_{+}^{3} := \sup\{P(f^{3}_{+}) : f \in \mathscr{F}\},$$
(2.4)

and

$$v^2 := \sigma^2 + 2\psi(\bar{E}) \text{ and } \beta^3_+ := m^3_+ + 3\rho(\bar{E}).$$
 (2.5)

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(i) Let  $\theta$  be a Bernoulli random variable with distribution  $\mathcal{B}_0(-v^2, 1)$  (defined by (1.2)). Let  $\theta_1, \ldots, \theta_n$  be n independent copies of  $\theta$  and let  $S_n := \theta_1 + \ldots + \theta_n$ . Then for any function  $\varphi \in \mathcal{G}^2$ ,

$$\mathbb{E}[\varphi(Z - \mathbb{E}[Z])] \le \mathbb{E}[\varphi(S_n)]. \tag{a}$$

(ii) Let  $\beta := \min\{\beta_+^3, v^2\}$ . Then for any function  $\varphi \in \mathcal{G}^3$ ,

$$\mathbb{E}[\varphi(Z - \mathbb{E}[Z])] \le \mathbb{E}[\varphi(\Gamma_{n(v^2 - \beta)} + \tilde{\Pi}_{n\beta})], \tag{b}$$

where  $\Gamma_{n(v^2-\beta)}$  and  $\Pi_{n\beta}$  are independent and respectively defined by Definition 1.1 (ii) and (iii).

**Remark 2.5.** Inequalities (*a*) and (*b*) of Theorem 2.4 are not easy to compare directly. To this end, we start by observing that Inequality (*a*) yields the following (see [1, Theorem 1.1]):

$$\mathbb{E}[\varphi(Z - \mathbb{E}[Z])] \le \mathbb{E}[\varphi(\tilde{\Pi}_{n\beta})] \quad \text{for any } \varphi \in \mathcal{G}^2.$$
 (a')

One can find in Pinelis [6] a thorough study of the comparison between the right-hand sides of (b) and (a') as well as with other classical bounds. Let us mention here some facts.

(i) If  $\beta = v^2$ , then the right-hand side of (b) is equal to the right-hand side of (a'). Thus, since  $\mathcal{G}^3 \subset \mathcal{G}^2$ , Inequality (b) is relevant with respect to (a) if and only if  $\beta_+^3 < v^2$  (see also the point (ii) below).

(ii) For any  $v^2 > 0$  and any  $\varepsilon \in \left]0, 1\right[$ ,

$$\mathbb{E}[\varphi(\Gamma_{(1-\varepsilon)v^2} + \tilde{\Pi}_{\varepsilon v^2})] \le \mathbb{E}[\varphi(\tilde{\Pi}_{v^2})] \quad \text{for any } \varphi \in \mathcal{G}^2.$$
(2.6)

#### **3** Proofs

The starting point of the proofs is a martingale decomposition of Z which we briefly recall. Firstly by virtue of the monotone convergence theorem, we can suppose that  $\mathscr{F}$  is a finite class of functions. Set  $\mathcal{F}_0 := \{\emptyset, \Omega\}$ ,  $\mathcal{F}_k := \sigma(X_1, \ldots, X_k)$  and  $\mathcal{F}_n^k := \sigma(X_1, \ldots, X_{k-1}, X_{k+1}, \ldots, X_n)$  for all  $k = 1, \ldots, n$ . Let  $\mathbb{E}_k$  (respectively  $\mathbb{E}_n^k$ ) denote the conditional expectation operator associated with  $\mathcal{F}_k$  (resp.  $\mathcal{F}_n^k$ ). Set also  $Z^{(k)} := \sup\{nP_n(f) - f(X_k) : f \in \mathscr{F}\}$  and  $Z_k := \mathbb{E}_k[Z]$ . Let us number the functions of the class  $\mathscr{F}$  and consider the random variables

$$\tau := \inf\{i > 0 : nP_n(f_i) = Z\} \text{ and } \tau_k := \inf\{i > 0 : nP_n(f_i) - f_i(X_k) = Z^{(k)}\}.$$

We notice that

$$f_{\tau_k}(X_k) \le Z - Z^{(k)} \le f_{\tau}(X_k).$$
 (3.1)

Set now  $\xi_k := \mathbb{E}_k[f_{\tau_k}(X_k)]$  and let  $r_k$  be the nonnegative random variable such that  $\xi_k + r_k = Z_k - \mathbb{E}_k[Z^{(k)}]$ . Since  $\mathbb{E}_n^k[f_{\tau_k}(X_k)] = P(f_{\tau_k})$  (see, for instance, Section 3.2 in [3, Lemma 3.10]), the centering assumption on the elements of  $\mathscr{F}$  ensures that  $\mathbb{E}_{k-1}[\xi_k] = 0$  and thus

$$Z - \mathbb{E}[Z] = \sum_{k=1}^{n} \Delta_k, \quad \text{where } \Delta_k := \xi_k + r_k - \mathbb{E}_{k-1}[r_k]. \tag{3.2}$$

The important point is that  $\mathbb{E}_{k-1}[r_k]$  is a corrective term which is essentially small. This is the statement of the following lemma:

**Lemma 3.1.** We have  $0 \le \mathbb{E}_{k-1}[r_k] \le E_{n-k+1} \le 1$  a.s.

The proof is based on a property of exchangeability of variables, proved in Marchina [3]. Since it is the fundamental tool of the paper, we give again the proof for sake of completeness.

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Proof of Lemma 3.1. Since  $f \leq 1$  for any  $f \in \mathscr{F}$ ,  $E_k \leq 1$  for any k = 1, ..., n. Moreover,  $r_k \geq 0$  which implies  $\mathbb{E}_{k-1}[r_k] \geq 0$ . Let us now prove the main part  $\mathbb{E}_{k-1}[r_k] \leq E_{n-k+1}$ . We start by the following property of exchangeability of variables.

**Lemma 3.2.** For any integer  $j \ge k$ ,  $\mathbb{E}_{k-1}[f_{\tau}(X_k)] = \mathbb{E}_{k-1}[f_{\tau}(X_j)]$  a.s.

Proof of Lemma 3.2. By the definition of the random variable  $\tau$ , for every permutation on n elements  $\sigma$ ,  $\tau(X_1, \ldots, X_n) = \tau \circ \sigma(X_1, \ldots, X_n)$  almost surely. Applying now this fact to  $\sigma = (k j)$  (the transposition which exchanges k and j), it suffices to use Fubini's theorem (recalling that  $j \ge k$ ) to complete the proof.

Hence,

$$\mathbb{E}_{k-1}[f_{\tau}(X_k)] = \mathbb{E}_{k-1}[f_{\tau}(X_k) + \ldots + f_{\tau}(X_n)]/(n-k+1)$$
  
$$\leq \mathbb{E}_{k-1} \sup_{f \in \mathscr{F}} \{f(X_k) + \ldots + f(X_n)\}/(n-k+1) = E_{n-k+1}.$$
(3.3)

Recalling that  $\mathbb{E}_{k-1}[\xi_k] = 0$ , (3.1) yields that  $\mathbb{E}_{k-1}[r_k] \leq \mathbb{E}_{k-1}[f_{\tau}(X_k)]$ , which combined with (3.3) end the proof of Lemma 3.1.

We are now in position to prove Theorem 2.1.

Proof of Theorem 2.1. Observe first that  $\Delta_k \leq 1$  by (3.1) and the uniform boundedness condition on  $\mathscr{F}$ . Then, in view of Proposition 1.3 (*i*), it remains us to bound up the conditional variance with respect to  $\mathcal{F}_{k-1}$  of  $\Delta_k$ . This is the subject of the following lemma:

**Lemma 3.3.** For any k = 1, ..., n,

$$\mathbb{E}_{k-1}[\Delta_k^2] \le \sup_{m \in [0, E_{n-k+1}]} (m+a)(1-m)$$
 a.s.

Proof of Lemma 3.3. A classical result due to Hoeffding [2] (see his Inequalities (4.1) and (4.2)) states that any bounded random variable X, such that  $a \le X \le b$  for some reals a and b, satisfies

$$\mathbb{E}[\varphi(X)] \le \mathbb{E}[\varphi(\theta)] \quad \text{for any convex function } \varphi, \tag{3.4}$$

where  $\theta$  is a Bernoulli random variable with distribution  $\mathcal{B}_{\mathbb{E}[X]}(a, b)$  (defined by (1.2)). In particular,  $\operatorname{Var}(X)$  is lower than  $\operatorname{Var}(\theta)$ . We apply this result conditionally to  $\mathcal{F}_{k-1}$  to the variable  $\xi_k + r_k$  which has its values in [-a, 1] by (3.1). Recalling now (1.3), one immediately obtains

$$\operatorname{Var}(\xi_k + r_k \mid \mathcal{F}_{k-1}) \le (\mathbb{E}_{k-1}[r_k] + a)(1 - \mathbb{E}_{k-1}[r_k]) \text{ a.s.},$$
(3.5)

which combined with Lemma 3.1 conclude the proof of Lemma 3.3.

Let us now complete the proof of Theorem 2.1. Define the function V by

$$V(m) := (m+a)(1-m) \text{ for any } m \in [0, E_{n-k+1}].$$
(3.6)

Case (i):  $a \ge 1$ . Since  $(1-a)/2 \le 0$ , V is decreasing on  $[0, E_{n-k+1}]$  and then by Lemma 3.3,  $\mathbb{E}_{k-1}[\Delta_k^2] \le a$ . Thus Inequality (a) follows by Proposition 1.3 (i).

Case (*ii*): a < 1. Here the maximum of V is reached at

$$m = \min(E_{n-k+1}, (1-a)/2)$$

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Thus Lemma 3.3 implies

$$\mathbb{E}_{k-1}[\Delta_k^2] \le (a+1)^2 \,\psi\Big(\frac{a+E_{n-k+1}}{a+1}\Big). \tag{3.7}$$

Furthermore, since  $\psi$  is concave,

$$\frac{1}{n}\sum_{k=1}^{n}\psi\Big(\frac{a+E_{n-k+1}}{a+1}\Big) \le \psi\Big(\frac{a+\bar{E}}{a+1}\Big). \tag{3.8}$$

Hence Inequality (b) follows again from Proposition 1.3 (i) together with (3.7)–(3.8). The proof of Theorem 2.1 is now complete.  $\Box$ 

*Proof of Theorem 2.4.* Case (*i*). We start from (3.2). Since  $\mathbb{E}_{k-1}[\xi_k] = 0$ , we get

$$\mathbb{E}_{k-1}[\Delta_k^2] = \mathbb{E}_{k-1}[(\xi_k + r_k)^2] - (\mathbb{E}_{k-1}[r_k])^2.$$
(3.9)

Since  $\mathbb{E}_n^k[f_{\tau_k}^2(X_k)] = P(f_{\tau_k}^2)$ , the conditional Jensen inequality implies that  $\mathbb{E}_{k-1}[\xi_k^2] \leq \sigma^2$ . Therefore, from (3.9) and the fact that (3.1) implies  $\xi_k + r_k \leq 1$ , we get

$$\mathbb{E}_{k-1}[\Delta_k^2] \leq \sigma^2 + \mathbb{E}_{k-1}[2\xi_k r_k + r_k^2] - (\mathbb{E}_{k-1}[r_k])^2 \\
\leq \sigma^2 + 2 \mathbb{E}_{k-1}[r_k] - (\mathbb{E}_{k-1}[r_k])^2 \\
\leq \sigma^2 + 2 \psi(\mathbb{E}_{k-1}[r_k]) \\
\leq \sigma^2 + 2 \psi(\mathbb{E}_{n-k+1}),$$
(3.10)

where the last inequality follows from Lemma 3.1 and  $\psi$  is the nondecreasing function already defined in (2.1). Then the same conclusion as in the proof of the Case (*ii*) of Theorem 2.1 allows us to conclude the proof of Inequality (*a*) of Theorem 2.4.

Case (ii). Since  $x \mapsto x_+^3$  is a convex function, we have

$$\Delta_{k+}^{3} \le \left(\xi_{k} - \mathbb{E}_{k-1}[r_{k}]\right)_{+}^{3} + 3r_{k}\left(\xi_{k} + r_{k} - \mathbb{E}_{k-1}[r_{k}]\right)_{+}^{2}.$$
(3.11)

In the same way as previously, since  $\mathbb{E}_n^k[f_{\tau_k+}^3(X_k)] = P(f_{\tau_k+}^3)$ , one has  $\mathbb{E}_{k-1}[\xi_{k+}^3] \le m_+^3$ . Therefrom, recalling that  $r_k \ge 0$  and  $\xi_k + r_k \le 1$ , we get

$$\mathbb{E}_{k-1}[\Delta_{k+1}^3] \le m_+^3 + 3\,\rho(\mathbb{E}_{k-1}[r_k]) \le m_+^3 + 3\,\rho(E_{n-k+1}),\tag{3.12}$$

where the last inequality follows from Lemma 3.1 and  $\rho$  is the nondecreasing function defined in (2.3). Since  $\rho$  is concave, we complete the proof in the same way as the Case (*i*) by using Proposition 1.3 (*ii*) in place of Proposition 1.3 (*i*).

#### References

- [1] V. Bentkus. (2004). On Hoeffding's inequalities, Ann. Probab. 32, no. 2, 1650–1673. MR-2060313
- [2] W. Hoeffding. (1963). Probability inequalities for sums of bounded random variables. J. Amer. Statist. Assoc. 58, 13–30.
- [3] A. Marchina. (2017). Concentration inequalities for suprema of unbounded empirical processes. Preprint on <hal-01545101>.
- [4] I. Pinelis. (1998). Optimal tail comparison based on comparison of moments. High dimensional probability, Springer, pp. 297–314. MR-1652335
- [5] I. Pinelis. (1999). Fractional sums and integrals of r-concave tails and applications to comparison probability inequalities. Advances in stochastic inequalities (Atlanta, GA, 1997) 234, 149–168. MR-1694770

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- [6] I. Pinelis. (2009). On the Bennett-Hoeffding inequality. Preprint. Available at arXiv:0902.4058 [math.PR].
- [7] I. Pinelis. (2014). On the Bennett-Hoeffding inequality. Ann. Inst. H. Poincaré Probab. Statist., 50, no. 1, 15–27. MR-3161520
- [8] I. Pinelis. (2016). Convex cones of generalized multiply monotone functions and the dual cones. Banach J. Math. Anal. 10, no. 4, 864–897. MR-3555754
- [9] E. Rio. (2001). Inégalités de concentration pour les processus empiriques de classes de parties. *Probab. Theory Related Fields* **119**, no. 2, 163–175.
- [10] A.W. van der Vaart and J.A. Wellner. (1996). Weak convergence and empirical processes: With applications to Statistics. Springer Series in Statistics. New-York: Springer. MR-1385671

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