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A CONCENTRATION INEQUALITY FOR THE SUM OF INDEPENDENT  
SYMMETRICALLY DISTRIBUTED RANDOM VARIABLES.

G. Kallianpur

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University of Minnesota\*  
Minneapolis, Minnesota

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Bounds on the Lévy concentration function [1] have been obtained by Kolmogorov ([2], [3]) and applied by him in [2]. A strengthened version of Kolmogorov's result has recently been given by B. A. Rogozin [4] who bases his proof on an inequality stated below. It is of interest in the study of central limit theorems (see [1]) to obtain bounds on the concentration function of the sum of independent symmetrically distributed random variables. We shall show how Rogozin's inequality can also be used to derive such a bound.

The concentration function of a random variable  $X$  is defined as

$$(1) \quad Q_X(a) = \max_{-\infty < x < \infty} \text{Prob} \left\{ x \leq X \leq x+a \right\}$$

Rogozin's inequality is the following:

Let  $X_k$  ( $k=1, \dots, n$ ) be independent random variables such that

$$P(X_k = x_k) = P(X_k = -x_k) = \frac{1}{2} \quad \text{where } x_k \geq l_k > 0.$$

If  $S = \sum_{k=1}^n X_k$  and  $L \geq \max_k l_k$ , then

$$(2) \quad Q_S(L) \leq C \cdot L \left( \sum_{k=1}^n l_k^2 \right)^{-\frac{1}{2}} \quad \text{when } C \text{ is a positive absolute constant.}$$

From the above inequality it is easy to deduce that (2) holds for independent symmetric random variables  $X_k$  whose probability mass is concentrated entirely outside the interval  $(-l_k, l_k)$ .

Lemma. Let  $X_k$  ( $k=1, \dots, n$ ) be independent random variables, symmetrically distributed about zero and such that  $|X_k| \geq l_k$  with probability one. Then for  $L \geq \max_k l_k$ ,  $Q_S(L)$  has the same bound as in (2).

To prove the lemma, observe that if we define  $T = \sum_{k=1}^n \epsilon_k Z_k$  where

the random variables  $\{\epsilon_k\}$ ,  $\{Z_k\}$  are mutually independent such that  $Z_k$  is distributed as  $|X_k|$  and  $\epsilon_k = +1$  or  $-1$  with equal probability, then  $\mathcal{L}(T)$ , the probability law or distribution of  $T$  equals  $\mathcal{L}(S)$ . Let  $z_k$  ( $k=1, \dots, n$ ) be fixed numbers with  $z_k \geq 1, z_k > 0$  and let  $T' = \sum_{k=1}^n \epsilon_k z_k$ .

We may then apply Rogozin's result to  $T'$  and obtain

$$Q_{T'}(L) \leq C L \left( \sum_{k=1}^n 1_k^2 \right)^{-\frac{1}{2}}.$$

An elementary argument now shows that  $Q_T(L)$  has also the same bound.

The lemma follows since  $Q_T(L) = Q_S(L)$ . We are now in a position to prove our main result.

Theorem. Let  $X_k$  ( $k=1, \dots, n$ ) be independent random variables symmetrically distributed about zero. If  $1_k$  are arbitrary positive numbers and  $L \geq \max_k 1_k$ , then.

$$(3) \quad Q_S(L) \leq AL \delta^{-\frac{1}{2}}, \quad \text{where } A \text{ is a positive absolute constant and}$$

$$\delta = \sum_{k=1}^n 1_k^2 P(|X_k| \geq 1_k).$$

The following corollaries are of interest in connection with the approximation by infinitely divisible distributions to the law of  $S$ .

Upon setting  $1_k=1$  in (3) we immediately obtain

$$\text{Corollary 1.} \quad Q_S(L) \leq \frac{A \cdot L}{L} \left\{ \sum_{k=1}^n P(|X_k| \geq 1) \right\}^{-\frac{1}{2}} \quad (L \geq 1).$$

Corollary 2. If the  $X_k$ 's are no longer assumed to be symmetric we have the following inequality:

$$(4) \quad Q_S(L) \leq C \cdot \left(\frac{L}{1}\right)^{\frac{1}{2}} \left[ \sum_{k=1}^n P(|X_k - m_k| \geq 21) \right]^{-\frac{1}{2}}, \quad (L \geq 1 > 0)$$

where  $C$  is a positive constant and  $m_k$  is the median of  $X_k$ .

In deducing Corollary 2 from Corollary 1 we make use of the easily verifiable facts that if  $\xi$  and  $\eta$  are independent, identically distributed random variables with median  $m$ , then

$$P(|\xi - \eta| \geq 1) \geq \frac{1}{2} P(|\xi - m| \geq 1) \quad \text{and} \quad Q_{\xi}^2(1) \leq Q_{\xi - \eta}(21).$$

In the case of symmetric random variables Kolmogorov's inequality follows from Corollary 1. Since  $P(|X_k| \geq 1) \geq 1 - Q_{X_k}(21)$  we obtain at once the inequality

$$Q_S(L) \leq A \cdot L^{-1} \left[ \sum_{k=1}^n \left\{ 1 - Q_{X_k}(21) \right\} \right]^{-\frac{1}{2}}.$$

In the general case the following inequality similarly follows from (4).

$$Q_S(L) \leq A \cdot L^{\frac{1}{2}} \cdot L^{-\frac{1}{2}} \left[ \sum_{k=1}^n \left\{ 1 - Q_{X_k}(21) \right\} \right]^{-\frac{1}{2}}.$$

We now give the proof of the theorem. Let  $\xi_k$  ( $k=1, \dots, n$ ) be independent random variables such that  $P(\xi_k=1) = 1 - P(\xi_k=0) = p_k$  where  $p_k = P(|X_k| \geq 1_k)$ . Also introduce independent random variables  $U_k$  and  $V_k$  such that the distribution of  $U_k$  is equal to the conditional distribution of  $X_k$  given  $|X_k| < 1_k$ , and the distribution of  $V_k$  is the conditional distribution of  $X_k$  given  $|X_k| \geq 1_k$ . If we set

$$Z = \sum_k \left\{ \xi_k V_k + (1 - \xi_k) U_k \right\}, \quad \text{then clearly we have } \mathcal{L}(Z) = \mathcal{L}(S).$$

Denote by  $E$  the event  $\left\{ \xi_{k_1}=1, \xi_k=0 \text{ if } k \neq k_1, i=1, \dots, v \text{ and} \right.$

$$\left. U_k = u_k \text{ for } k=1, \dots, n \right\}; \quad \text{setting } a = \sum_{k \neq \text{any } k_1} u_k \quad \text{we find that the}$$

conditional probability

$$P[ x \leq Z \leq x+L | E ] = P[ x-a \leq \sum_{i=1}^v V_{k_i} \leq x-a+L ].$$

Hence

$$P[ x \leq Z \leq x+L ] = \int P[ x-a \leq \sum_{i=1}^v V_{k_i} \leq x-a+L ] dF$$

where  $F$  is the joint distribution of  $\{U_k\}$  and  $\{\xi_k\}$  and the integration

is over all possible values  $\{u_k\}$  of the  $U_k$  and of the variables  $\{\xi_k\}$ .

We bound the integrand by  $C \cdot L \left[ \sum_{i=1}^v 1_{k_i}^2 \right]^{-\frac{1}{2}}$  over the region  $\sum_k 1_k^2 \xi_k^2 > \frac{\delta}{2}$

(which we may do on account of the lemma proved above) and by 1 over the

complementary region  $\sum_k 1_k^2 \xi_k^2 \leq \delta/2$ . We then obtain

$$Q_Z(L) \leq C L \left(\frac{\delta}{2}\right)^{-\frac{1}{2}} + P\left[\sum_{k=1}^n 1_k^2 \xi_k^2 \leq \frac{\delta}{2}\right].$$

Since, by definition,  $\delta = E\left(\sum_k 1_k^2 \xi_k^2\right)$  and since  $\text{Var}\left(\sum_k 1_k^2 \xi_k^2\right) \leq L^2 \delta$

we have  $P\left[\sum_k 1_k^2 \xi_k^2 \leq \frac{\delta}{2}\right] \leq P\left[\left|\sum_k 1_k^2 \xi_k^2 - E\left(\sum_k 1_k^2 \xi_k^2\right)\right| \geq \frac{\delta}{2}\right]$   
 $\leq 4L^2 \delta^{-1}$ .

Hence

$$(5) \quad Q_Z(L) \leq \sqrt{2C} L \delta^{-\frac{1}{2}} + 4L^2 \delta^{-1}.$$

The inequality  $Q_Z(L) \leq (\sqrt{2C} + 4) L \delta^{-\frac{1}{2}}$  now follows from (5)

if  $L \delta^{-\frac{1}{2}} \leq 1$  and is trivially satisfied if  $L \delta^{-\frac{1}{2}} \geq 1$ . This

completes the proof of the theorem.

### References

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