

## On a result of A. Jakubowski

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Chebyshev's other inequality is the statement that if  $f$  and  $g$  are non-decreasing functions from  $\mathbf{R}$  to  $\mathbf{R}$  and if  $X$  is a random variable such that both  $f(X)$  and  $g(X)$  are square integrable, then  $\text{Cov}[f(X), g(X)] \geq 0$ . Jakubowski [J] has recently discussed the equality case, showing that if  $\text{Cov}[f(X), g(X)] = 0$  then either  $f(X)$  is constant a.s. or  $g(X)$  is constant a.s. I record here another proof of this assertion.

The neatest proof of the inequality goes as follows. Let  $Y$  be an independent copy of  $X$ , and notice that

$$(1) \quad [f(X) - f(Y)] \cdot [g(X) - g(Y)] \geq 0$$

everywhere, and so

$$(2) \quad \text{Cov}[f(X), g(X)] = \frac{1}{2} \mathbf{E} [[f(X) - f(Y)] \cdot [g(X) - g(Y)]] \geq 0.$$

As noted in [J], this approach also gives insight into the case in which  $f(X)$  and  $g(X)$  are uncorrelated. For if  $\text{Cov}[f(X), g(X)] = 0$  then (1) and (2) imply that

$$(3) \quad [f(X) - f(Y)] \cdot [g(X) - g(Y)] = 0, \quad \text{a.s.}$$

We now argue that if  $f(X)$  and  $g(X)$  are uncorrelated, and  $\text{Var}[f(X)] > 0$ , then  $\text{Var}[g(X)] = 0$ .

First notice that by the special case  $f = g$  in (2),

$$(4) \quad \mathbf{E} [[f(X) - f(Y)]^2] = 2\text{Var}[f(X)] > 0.$$

Therefore  $\mathbf{P}[f(X) < f(Y)] = \frac{1}{2} \mathbf{P}[f(X) \neq f(Y)] > 0$ . Let  $A_0 := \{f(X) < f(Y)\}$ . In view of (3) this differs from  $A := \{f(X) < f(Y), g(X) = g(Y)\}$  by a null set.

Now introduce a third random variable  $Z$  independent of  $(X, Y)$  with the same distribution as  $X$  and  $Y$ . Consider the event

$$(5) \quad B := A \cap \{g(Y) < g(Z)\}.$$

On  $B$  we have  $g(X) < g(Z)$ ; by (3) stated for the pair  $(X, Z)$  we have  $f(X) = f(Z)$  a.s. on  $B$ . Then, by the monotonicity of  $f$ ,  $f(X) = f(Y)$  a.s. on  $B$ . It follows that  $\mathbf{P}[B] = 0$ . Similarly,  $\mathbf{P}[A \cap \{g(Z) < g(X)\}] = 0$ . Thus,

$$(6) \quad g(X) = g(Y) = g(Z), \quad \text{a.s. on } A.$$

Let  $\mu$  denote the common distribution of  $X, Y, Z$ . From (6) and the fact that  $Z$  is independent of  $A$ , we deduce that

$$(7) \quad g(X) = g(Y) = g(z), \quad \text{a.s. on } A,$$

for  $\mu$ -a.e.  $z \in \mathbf{R}$ . Let  $J$  denote the smallest interval carrying all the mass of  $\mu$ . Then because  $g$  is monotone (7) implies that

$$(8) \quad g(X) = g(Y) = g(z), \forall z \in J, \quad \text{a.s. on } A.$$

But  $\mathbf{P}[A] > 0$  so  $A$  has at least one element. Consequently, (8) implies that  $g$  is constant on  $J$ , and therefore  $g(X)$  is constant a.s.  $\square$

## References

- [J] Jakubowski, A.: A complement to the Chebyshev integral inequality, *Stat. Probab. Letters* **168** (2021) 108934.