journal of inequalities in pure and applied mathematics
http://jipam.vu.edu.au
issn: 1443-5756
Volume 9 (2008), Issue 1, Article 3, 8 pp.

AN INEQUALITY FOR CORRELATED MEASURABLE FUNCTIONS

FABIO ZUCCA
DIPARTIMENTO DI MATEMATICA
POLITECNICO DI MILANO
PIAZZA LEONARDO DA VINCI 32
20133 MILANO, ITALY.
fabio.zucca@polimi.it
URL: http://www1.mate.polimi.it/~zucca

Received 01 December, 2007; accepted 24 January, 2008
Communicated by S. Abramovich

ABSTRACT. A classical inequality, which is known for families of monotone functions, is generalized to a larger class of families of measurable functions. Moreover we characterize all the families of functions for which the equality holds. We give two applications of this result, one of them to a problem arising from probability theory.

Key words and phrases: Integral inequalities, Measure, Cartesian product, Ordered set.

2000 *Mathematics Subject Classification.* 26D15, 28A25.

1. INTRODUCTION

The aim of this paper is to generalize an inequality, originally due to Chebyshev and then rediscovered by Stein in [4]. Usually this result is stated for monotonic real functions: the classical inequality is

$$(b-a)\int_a^b f(x)g(x)\mathrm{d}x \geq \int_a^b f(x)\mathrm{d}x \int_a^b g(x)\mathrm{d}x$$

where f and g are monotonic (in the same sense) real functions (see for instance [4], [3] and [2] for a more general version). If $a = b - 1$ then this inequality has a probabilistic interpretation, namely $\mathbb{E}[fg] - \mathbb{E}[f]\mathbb{E}[g] \geq 0$ (where $\mathbb{E}$ denotes the expectation), that is, the covariance of f and g is nonnegative.

Our approach allows us to prove the inequality for functions defined on a general measurable space, hence we go beyond the usual ordered set $\mathbb{R}$. More precisely, we prove an analogous result for general families of measurable functions that we call correlated functions (see Definition 2.1 for details). In particular, we characterize all the families of functions for which the equality holds.

Here is the outline of the paper. In Section 2 we introduce the terminology and the main tools needed in the sequel. In particular, Sections 2.1 and 2.2 are devoted to the construction of an

The author thanks S. Mortola for useful discussions.
355-07

order relation and a σ-algebra on a particular quotient space. In Section 3 we state and prove our main result (Theorem 3.1) which involves k correlated functions; the special case $k = 2$ requires weaker assumptions (see also Remark 3.4). We give two applications of this inequality in Section 4: the first one involves a particular class of power series, while the second one comes from probability theory.

2. PRELIMINARIES AND BASIC CONSTRUCTIONS

We start from a very general setting. Let us consider a set X, a partially ordered space $(Y, \geq_Y)$ and a family $\mathcal{N} = \{f_i\}_{i\in\Gamma}$ (where Γ is an arbitrary set) of functions in Y^X. We consider the equivalence relation on X

$$x \sim y \Longleftrightarrow f_i(x) = f_i(y), \quad \forall i \in \Gamma$$

and we denote by $X/_\sim$ the quotient space, by $[x]$ the equivalence class of $x \in X$ and by π the natural projection of X onto $X/_\sim$. Roughly speaking, by means of this procedure, we identify points in X which are not separated by the family $\mathcal{N}$.

To the family $\mathcal{N}$ corresponds a natural counterpart $\mathcal{N}_\sim = \{\phi_{f_i}\}_{i\in\Gamma}$ of functions in $Y^{X/\sim}$, where, by definition, $\phi_f([x]) := f(x)$, for all $x \in X$ and for every $f \in Y^X$ satisfying

$$\forall x, y \in X : x \sim y \Longrightarrow f(x) = f(y) \tag{2.1}$$

(this holds in particular for all the functions in $\mathcal{N}$). It is clear that the family $\mathcal{N}_\sim$ separates the points of $X/_\sim$.

Given any function g defined on $X/_\sim$ we denote by π_g the function $g \circ \pi$; observe that $\phi_{\pi_g} = g$ for all $g \in Y^{X/\sim}$ and $\pi_{\phi_f} = f$ for every f satisfying equation (2.1). Clearly $g \mapsto \pi_g$ is a bijection from $Y^{X/\sim}$ onto the subset of a function in Y^X satisfying equation (2.1).

Note that given $f, f_1 \in Y^X$ which satisfy equation (2.1) (resp. $g, g_1 \in Y^{X/\sim}$) then $f \geq_Y f_1$ (resp. $g \geq_Y g_1$) implies $\phi_f \geq \phi_{f_1}$ (resp. $\pi_g \geq \pi_{g_1}$).

2.1. Induced order.

In order to prove Theorem 3.1 we cannot take advantage, as in the classical formulation, of an order relation on the set X. Under some reasonable assumptions (see Definition 2.1 below) we can transfer the order relation from Y to $X/_\sim$ where we already defined a family $\mathcal{N}_\sim$ related to the original $\mathcal{N}$. This will be enough for our purposes.

Definition 2.1. The functions in $\mathcal{N}$ are *correlated if, for all* $i \in \Gamma$ *and* $x, y \in X$,

$$f_i(x) >_Y f_i(y) \Longrightarrow f_j(x) \geq_Y f_j(y), \quad \forall j \in \Gamma. \tag{2.2}$$

We note that the definition above can be equivalently stated as follows: for all $i, j \in \Gamma$ and $x \in X$,

$$f_i^{-1}((-\infty, f_i(x))) \subseteq f_j^{-1}((-\infty, f_j(x)]).$$

Besides, if $Y = \mathbb{R}$ with its natural order, then the functions in $\mathcal{N}$ are correlated if and only if for all $i, j \in \Gamma$ and $x, y \in X$,

$$(f_i(x) - f_i(y))(f_j(x) - f_j(y)) \geq 0. \tag{2.3}$$

In particular if X is a totally ordered set and all the functions in $\mathcal{N}$ are nondecreasing (or nonincreasing) then they are correlated.

A family of correlated functions induces a natural order relation on the quotient space $X/_\sim$.

Lemma 2.1. *If the functions in* $\mathcal{N}$ *are correlated then the relation on* $X/_\sim$

$$[x] \geq_\sim [y] \Longleftrightarrow f_i(x) \geq_Y f_i(y), \quad \forall i \in \Gamma$$

is a partial order. If $(Y, \geq_Y)$ *is a totally ordered space then the same holds for* $(X/_\sim, \geq_\sim)$. *Moreover* $\mathcal{N}_\sim$ *is a family of nondecreasing functions (hence they are correlated).*

Proof. It is straightforward to show that $\geq_\sim$ is a well-defined partial order (clearly it does not depend on the choice of x (and y) within an equivalence class). We prove that, if $\geq_Y$ is a total order, the same holds for $\geq_\sim$. Indeed if $[x] \neq [y]$ then there exists $i \in \Gamma$ such that $f_i(x) \neq f_i(y)$; suppose that $f_i(x) > f_i(y)$ then, by equation (2.2), $[x] >_\sim [y]$. It is trivial to prove that ϕ_{f_i} is nondecreasing for every $i \in \Gamma$, whence they are correlated since the space $(X/_\sim, \geq_\sim)$ is totally ordered. □

A subset I of an ordered set, say Y, is called an interval if and only if for all $x, y \in I$ and $z \in Y$ then $x \geq_Y z \geq_Y y$ implies $z \in I$. Note that given an interval $I \subseteq Y$ then $\phi_{f_i}^{-1}(I)$ is an interval of $X/_\sim$ for every $i \in \Gamma$.

Given $x, y \in X$ such that $[x] \geq_\sim [y]$ we define the interval $[[y], [x]) := \{[z] \in X/_\sim : [y] \leq [z] < [x]\}$; the intervals $[[y], [x]]$, $([y], [x]]$ and $([y], [x])$ are defined analogously. In particular, for any $x \in X$, we denote by $[[x], +\infty)$ and $(-\infty, [x]]$ the intervals $\{[y] \in X/_\sim : [y] \geq_\sim [x]\}$ and $\{[y] \in X/_\sim : [x] \geq_\sim [y]\}$ respectively.

2.2. **Induced σ-algebra and measure.** This construction can be carried on under general assumptions. Let us consider a measurable space with a positive measure (X, Σ_X, μ) and an equivalence relation $\sim$ on X such that for all $x \in X$ and $A \in \Sigma_X$,

$$x \in A \Longrightarrow [x] \subseteq A. \tag{2.4}$$

There is a natural way to construct a σ-algebra on $X/_\sim$, namely define

$$\Sigma_\sim := \{\pi(A) : A \in \Sigma_X\}$$

where $\pi(A) := \{[x] : x \in A\}$. This is the largest σ-algebra on $X/_\sim$ such that the projection map π is measurable. Observe that $A \mapsto \pi(A)$ is a bijection from Σ_X onto $\Sigma_\sim$. It is natural to define a measure $\overline{\mu} := \mu_\pi$ by

$$\overline{\mu}(\pi(A)) = \mu(A), \quad \forall A \in \Sigma_X.$$

It is well known that a function $g : X/_\sim \to \mathbb{R}$ is measurable if and only if π_g is measurable. Moreover, g is integrable (with respect to $\overline{\mu}$) if and only if π_g is integrable (with respect to μ) and

$$\int_X \pi_g \mathrm{d}\mu = \int_{X/_\sim} g \mathrm{d}\overline{\mu}. \tag{2.5}$$

We say that a function g is *integrable* if at least one of the integrals of the two nonnegative functions $g^+ := \max(g, 0)$ and $g^- := -\min(g, 0)$ is finite; hence the integral of g can be unambiguously defined as the difference of the two integrals (where $\pm\infty + z := \pm\infty$ for all $z \in \mathbb{R}$ and $0 \cdot \pm\infty := 0$). This notion is slightly weaker than the usual one: to remark the difference, when the integrals of g^+ and g^- are both finite the function g is called *summable*.

It is a simple exercise to check that the equivalence relation defined in Section 2.1 satisfies equation (2.4) if $\Sigma_X = \sigma(f_i : i \in \Gamma)$ (that is, Σ_X is the minimal σ-algebra such that all the functions in $\mathcal{N}$ are measurable); this equivalence relation along with its induced σ-algebra and measure will play a key role in the next section.

Remark 2.2. It is easy to show that if $h, r : X \mapsto \mathbb{R}$ are two integrable functions such that the sum $\int_X h \mathrm{d}\mu + \int_X r \mathrm{d}\mu$ is not ambiguous (i.e., it is not true that $\int_X h \mathrm{d}\mu = \pm\infty$ and $\int_X r \mathrm{d}\mu = \mp\infty$), then $h + r$ is integrable and

$$\int_X (h + r) \mathrm{d}\mu = \int_X h \mathrm{d}\mu + \int_X r \mathrm{d}\mu \tag{2.6}$$

(both sides possibly being equal to $\pm\infty$). This will be useful in the proof of Lemma 3.3.

3. MAIN RESULT

Throughout this section we consider a measurable space with finite positive measure (X, Σ_X, μ) and a family of correlated functions $\mathcal{N} = \{f_i\}_{i\in\Gamma}$, where $\Sigma_X = \sigma(f_i : i \in \Gamma)$. Let us consider $Y = \mathbb{R}$ with its natural order $\geq$. The equivalence relation $\sim$, the (total) order $\geq_\sim$ and the space $(X/_\sim, \Sigma_\sim, \overline{\mu})$ are introduced according to Sections 2.1 and 2.2. It is clear that $\Sigma_\sim$ contains the σ-algebra generated by the set of intervals $\{\phi_{f_i}^{-1}(I) : i \in \Gamma, I \subseteq \mathbb{R} \text{ is an interval}\}$. More precisely, it is easy to see that, by construction, all the intervals of the totally ordered set $(X/_\sim, \geq_\sim)$ are measurable since $\mathcal{N}_\sim$ separates points.

The main result is the following.

Theorem 3.1. *Let $\mu(X) < +\infty$.*

(1) *If f, g are two integrable, μ-a.e. correlated functions such that fg is integrable then*

$$\mu(X)\int_X fg\mathrm{d}\mu \geq \int_X f\mathrm{d}\mu \int_X g\mathrm{d}\mu. \tag{3.1}$$

Moreover, if f, g are summable, then in the previous equation the equality holds if and only if at least one of the functions is μ-a.e constant.

(2) *If $\{f_i\}_{i=1}^k$ is a family of measurable functions on X which are nonnegative and μ-a.e. correlated, then*

$$\mu(X)^{k-1}\int_X \prod_{i=1}^k f_i\mathrm{d}\mu \geq \prod_{i=1}^k \int_X f_i\mathrm{d}\mu. \tag{3.2}$$

Moreover, if $\int_X f_i\mathrm{d}\mu \in (0, +\infty)$ for all $i = 1, \ldots, k$, then in the previous equation the equality holds if and only if at least $k-1$ functions are μ-a.e. constant.

Before proving this theorem, let us warm up with the following lemma; though it will not be used in the proof of Theorem 3.1, nevertheless it sheds some light on the next step.

Lemma 3.2. *Let $\mathcal{N} := \{\{x_i(j)\}_{i\in\mathbb{N}}\}_{j=1}^k$ be a family of nonnegative and nondecreasing sequences and $\{\mu_i\}_{i\in\mathbb{N}}$ be a family of strictly positive real numbers. If $\sum_i \mu_i < +\infty$ then*

$$\left(\sum_i \mu_i\right)^{k-1} \sum_i \prod_{j=1}^k x_i(j)\mu_i \geq \prod_{j=1}^k \sum_i x_i(j)\mu_i. \tag{3.3}$$

Moreover, if for every j we have $0 < \sum_i x_i(j) < +\infty$, then the equality holds if and only if at least $k-1$ sequences are constant.

Proof. We prove the first part of the claim for two finite sequences $\{x_i\}_{i=1}^n$ and $\{y_i\}_{i=1}^n$, since the general case follows easily by induction on k and using the Monotone Convergence Theorem as n tends to infinity.

It is easy to prove that

$$\begin{aligned}\sum_{i=1}^n \mu_i \sum_{i=1}^n x_i y_i \mu_i - \sum_{i=1}^n x_i\mu_i \sum_{i=1}^n y_i\mu_i &= \sum_{i,j:i\geq j} (x_i - x_j)(y_i - y_j)\mu_i\mu_j \\ &= \sum_{i,j:i>j} (x_i - x_j)(y_i - y_j)\mu_i\mu_j.\end{aligned} \tag{3.4}$$

Indeed,

$$\sum_{i=1}^n \mu_i \sum_{i=1}^n x_i y_i \mu_i = \sum_{i,j:i>j} (x_i y_i + x_j y_j)\mu_i\mu_j + \sum_{i=1}^n x_i y_i \mu_i^2$$

and

$$\sum_{i=1}^{n} x_i\mu_i \sum_{i=1}^{n} y_i\mu_i = \sum_{i,j:i>j} (x_iy_j + x_jy_i)\mu_i\mu_j + \sum_{i=1}^{n} x_iy_i\mu_i^2.$$

This implies easily that

$$\sum_{i=1}^{n} \mu_i \sum_{i=1}^{n} x_iy_i\mu_i - \sum_{i=1}^{n} x_i\mu_i \sum_{i=1}^{n} y_i\mu_i \geq 0.$$

If either at least $k-1$ sequences are constant or one sequence is equal to 0, then we have an equality. The same is true if $\sum_i x_i(j)\mu_i = +\infty$ for some j and $\sum_i x_i(j)\mu_i > 0$ for all j, since both sides of equation (3.3) are equal to $+\infty$. On the other hand, by using the first part of the theorem and by taking the limit in equation (3.4) as n tends to infinity, for all $1 \leq j_1 < j_2 \leq k$,

$$\begin{aligned}(3.5)\quad \left(\sum_i \mu_i\right)^{k-1} &\sum_i \prod_{j=1}^{k} x_i(j)\mu_i - \prod_{j=1}^{k} \sum_i x_i(j)\mu_i \\ &\geq \left(\sum_i \mu_i\right) \sum_i x_i(j_1)x_i(j_2)\mu_i \prod_{j\neq j_1,j_2} \sum_i x_i(j)\mu_i - \prod_{j=1}^{k} \sum_i x_i(j)\mu_i \\ &= \left(\prod_{j\neq j_1,j_2} \sum_i x_i(j)\mu_i\right) \sum_{i,i_1:i>i_1} (x_i(j_1) - x_{i_1}(j_1))(x_i(j_2) - x_{i_1}(j_2))\mu_i\mu_{i_1}.\end{aligned}$$

If both $\{x_i(j_1)\}_i$ and $\{x_i(j_2)\}_i$ are nonconstant, then there exist $r < l$ and $r_1 < l_1$ such that $x_r(j_1) < x_l(j_1)$ and $x_{r_1}(j_2) < x_{l_1}(j_2)$. This implies that $x_{\max(l,l_1)}(j_1) - x_{\min(r,r_1)}(j_1) > 0$ and $x_{\max(l,l_1)}(j_2) - x_{\min(r,r_1)}(j_2) > 0$, thus the right hand side of equation (3.5) is strictly positive (just consider the summation over $\{i, i_1 : i \geq \max(l, l_1), i_1 \leq \min(r, r_1)\}$) and we have a strict inequality in equation (3.3). □

The proof of the previous lemma clearly suggests a second lemma which will be needed in the proof of Theorem 3.1.

Lemma 3.3. *Let $\mathcal{N} := \{f, g\}$ where $f, g : X \to \mathbb{R}$ are two summable functions such that fg is integrable (for instance if f and g are μ-a.e. correlated). If $\mu(X) < +\infty$ then*

$$\begin{aligned}(3.6)\quad \mu(X)\int_X f(x)g(x)\mathrm{d}\mu(x) &= \int_X f(x)\mathrm{d}\mu(x)\int_X g(x)\mathrm{d}\mu(x) \\ &\quad + \frac{1}{2}\int_{X\times X} (f(x) - f(y))(g(x) - g(y))\mathrm{d}\mu(x)\mathrm{d}\mu(y).\end{aligned}$$

Proof. Note that

$$(3.7)\qquad f(x)g(x) + f(y)g(y) = f(x)g(y) + f(y)g(x) + (f(x) - f(y))(g(x) - g(y));$$

where $f(x)g(y)$ and $f(y)g(x)$ are summable on $X \times X$, since f, g are summable. If we define $h(x, y) := f(x)g(y) + f(y)g(x)$ and $r(x, y) := (f(x) - f(y))(g(x) - g(y))$ then, according to Remark 2.2, we just need to prove that h and r are integrable (since $h + r$ is integrable by hypothesis).

If f, g are summable then, by equation (3.7), fg is integrable if and only if $(f(x)-f(y))(g(x)-g(y))$ is integrable on $X \times X$ (since the sum of a summable function and an integrable function is an integrable function) and equation (3.6) follows. Clearly, if f and g are correlated, then $(f(x) - f(y))(g(x) - g(y))$ is nonnegative thus integrable. □

Proof of Theorem 3.1.

(1) By equation (2.5) it is enough to prove that

$$\overline{\mu}(X/_\sim)\int_{X/_\sim}\phi_f\phi_g\mathrm{d}\overline{\mu}\geq\int_{X/_\sim}\phi_f\mathrm{d}\overline{\mu}\int_{X/_\sim}\phi_g\mathrm{d}\overline{\mu}.$$

If f and g are summable then the claim follows from equation (3.6) of Lemma 3.3. Otherwise, without loss of generality, we may suppose that $\int_{X/_\sim}\phi_f\mathrm{d}\overline{\mu}\equiv\int_X f\mathrm{d}\mu=+\infty$. If $\int_{X/_\sim}\phi_g\mathrm{d}\overline{\mu}\equiv\int_X g\mathrm{d}\mu<0$, then there is nothing to prove. If $\int_X g\mathrm{d}\mu\geq 0$, then either $g=0$ μ-a.e., in this case, both sides of equation (3.1) are equal to 0, or there exists $x\in X/_\sim$ such that $\overline{\mu}([x,+\infty))>0$ and $\phi_f,\phi_g>0$ on $[x,+\infty)$ (since ϕ_f and ϕ_g are nondecreasing). Clearly, $\int_{[x,+\infty)}\phi_f\mathrm{d}\overline{\mu}=+\infty$ and $\phi_f(y)\phi_g(y)\geq\phi_f(y)\phi_g(x)$ for all $y\in[x,+\infty)$, hence both sides of equation (3.1) are equal to $+\infty$.

If one of the two functions is constant, then the equality holds. If f and g are nonconstant (that is, ϕ_f and ϕ_g are nonconstant), then there exist $x_0,y_0\in X/_\sim$ such that $x_0>_\sim y_0$, $\phi_f(x_0)>\phi_f(y_0)$, $\phi_g(x_0)>\phi_g(y_0)$, $\overline{\mu}((-\infty,y_0])>0$ and $\overline{\mu}([x_0,+\infty))>0$ (this can be done as in Lemma 3.3). Hence, using equation (3.6), we have that,

$$\begin{aligned}&\overline{\mu}(X/_\sim)\int_{X/_\sim}\phi_f\phi_g\mathrm{d}\overline{\mu}-\int_{X/_\sim}\phi_f\mathrm{d}\overline{\mu}\int_{X/_\sim}\phi_g\mathrm{d}\overline{\mu}\\&\geq\int_{[x_0,+\infty)\times(-\infty,y_0]}(\phi_f(x)-\phi_f(y))(\phi_g(x)-\phi_g(y))\mathrm{d}\overline{\mu}(x)\mathrm{d}\overline{\mu}(y)\\&\geq\overline{\mu}((-\infty,y_0])\,\overline{\mu}([x_0,+\infty))(\phi_f(x_0)-\phi_f(y_0))(\phi_g(x_0)-\phi_g(y_0))>0.\end{aligned}$$

(2) Let us suppose that f_i is summable for all $i=1,\dots,k$. It is enough to prove that

$$\overline{\mu}(X/_\sim)^{k-1}\int_{X/_\sim}\prod_{i=1}^k\phi_{f_i}\mathrm{d}\overline{\mu}\geq\prod_{i=1}^k\int_{X/_\sim}\phi_{f_i}\mathrm{d}\overline{\mu}.$$

In the previous part of the theorem, we proved the claim for two functions ϕ_f and ϕ_g; as in Lemma 3.2, the general case follows by induction on k.

If at least two functions are nonconstant, say ϕ_{f_1}, ϕ_{f_2}, then as before we may find $x_0,y_0\in X/_\sim$ such that $x_0>_\sim y_0$, $\phi_{f_1}(x_0)>\phi_{f_1}(y_0)$, $\phi_{f_2}(x_0)>\phi_{f_2}(y_0)$, $\overline{\mu}((-\infty,y_0])>0$ and $\overline{\mu}([x_0,+\infty))>0$ (this can be done as in Lemma 3.3). By applying the first part of the claim to the family (of $k-1$ functions) $\phi_{f_1},\phi_{f_2},\phi_{f_3},\dots,\phi_{f_k}$ (which are clearly still correlated since they are nondecreasing) and using equation (3.6) we have that,

$$\begin{aligned}&\overline{\mu}(X/_\sim)^{k-1}\int_{X/_\sim}\prod_{i=1}^k\phi_{f_i}\mathrm{d}\overline{\mu}-\prod_{i=1}^k\int_{X/_\sim}\phi_{f_i}\mathrm{d}\overline{\mu}\\&=\left(\mu(X/_\sim)\int_{X/_\sim}\phi_{f_1}\phi_{f_2}\mathrm{d}\overline{\mu}-\int_{X/_\sim}\phi_{f_1}\mathrm{d}\overline{\mu}\cdot\int_{X/_\sim}\phi_{f_2}\mathrm{d}\overline{\mu}\right)\prod_{i=3}^k\int_{X/_\sim}\phi_{f_i}\mathrm{d}\overline{\mu}\\&\geq\left(\int_{[x_0,+\infty)\times(-\infty,y_0]}(\phi_{f_1}(x)-\phi_{f_1}(y))(\phi_{f_2}(x)-\phi_{f_2}(y))\mathrm{d}\overline{\mu}(x)\mathrm{d}\overline{\mu}(y)\right)\prod_{i=3}^k\int_{X/_\sim}\phi_{f_i}\mathrm{d}\overline{\mu}\\&\geq\overline{\mu}((-\infty,y_0])\,\overline{\mu}([x_0,+\infty))(\phi_{f_1}(x_0)-\phi_{f_1}(y_0))(\phi_{f_2}(x_0)-\phi_{f_2}(y_0))\prod_{i=3}^k\int_{X/_\sim}\phi_{f_i}\mathrm{d}\overline{\mu}\\&>0\end{aligned}$$

since $0 < \int_{X/\sim} \phi_{f_i} \mathrm{d}\overline{\mu} < +\infty$ for all $i = 1, \ldots, k$, thus the second part of the claim is proved.

□

Note that if $\int_X f_i \mathrm{d}\mu = +\infty$ for some i and $\int_X f_j \mathrm{d}\mu > 0$ for all j (otherwise both sides of equation (3.2) are equal to 0) then both sides of equation (3.2) are equal to $+\infty$; indeed, apply the first part of the theorem to the family of correlated bounded functions $\{\min(f_i, n)\}_{i=1}^k$ (where $n \in \mathbb{N}$) and take the limit of both sides of equation (3.2) as n tends to $+\infty$.

Remark 3.4. According to Theorem 3.1, there is a difference between the case $k = 2$ and $k > 2$; indeed, in the latter case the inequality cannot be proved for integrable (or even summable) μ-a.e. correlated functions which are not nonnegative. Something happens in the inductive process, namely if $\{f_i\}_{i=1}^k$ are correlated this may not be true for $\{f_1 f_2, f_3, \ldots, f_k\}$ (if the functions are not positive). Here is a counterexample: take $X = [-1, 1]$ endowed with the Lebesgue measure, $f_1(x) = f_2(x) := x\mathbf{1}_{[-1,0]}(x)$ and $f_i(x) := x - f_1(x)$ for all $i \geq 3$.

Strictly speaking, Theorem 3.1 could be proved without the constructions of Sections 2.1 and 2.2; one has just to use carefully equation (2.3) and Lemma 3.3. Our approach simplifies the proof of Theorem 3.1 and gives a better understanding of the role of the correlation hypothesis (compared to the usual monotonicity).

We finally observe that if we consider two integrable *anticorrelated* functions (meaning that $(f(x) - f(y))(g(x) - g(y)) \leq 0$ for all $x, y \in X$) such that fg is integrable then, clearly, we have $\mu(X) \int_X fg \mathrm{d}\mu \leq \int_X f \mathrm{d}\mu \int_X g \mathrm{d}\mu$.

4. FINAL REMARKS AND EXAMPLES

Let us apply Theorem 3 to a class of power series. We consider $f(z) := \sum_{n=0}^{+\infty} a_n z^n$, where $\{a_n\}_n$ is a sequence of nonnegative real numbers and we suppose that $\{\rho^n a_n\}$ is nonincreasing (resp. nondecreasing) for some ρ such that $0 < \rho \leq R$ (where R is the radius of convergence). Then the function $z \mapsto (\rho - z) f(z)$ is nonincreasing (resp. nondecreasing) on $[0, \rho)$.

Indeed, if we suppose that $\{\rho^n a_n\}$ is nonincreasing then, for all z, γ such that $0 \leq z < \gamma < \rho$, we have

$$\begin{aligned}\sum_{n=0}^{+\infty} a_n z^n &= \sum_{n=0}^{+\infty} a_n \rho^n (z/\gamma)^n (\gamma/\rho)^n \\ &\geq \frac{\sum_{n=0}^{+\infty} a_n \gamma^n}{\sum_{n=0}^{+\infty} (\gamma/\rho)^n} \sum_{n=0}^{+\infty} (z/\rho)^n = \sum_{n=0}^{+\infty} a_n \gamma^n \frac{\rho - \gamma}{\rho - z},\end{aligned}$$

where, in the first inequality, we applied Theorem 3.1 to the (correlated) functions $f_1(n) := a_n \rho^n$ and $f_2(n) := (z/\gamma)^n$ defined on $\mathbb{N}$ endowed with the measure $\mu(A) := \sum_{n \in A} (\gamma/\rho)^n$. The case when $\{\rho^n a_n\}$ is nondecreasing is analogous (observe that now the functions f_1 and f_2 are anticorrelated). If $z < \rho < R$, then f_1 and f_2 are nonconstant functions, hence the function $z \mapsto (\rho - z) f(z)$ is strictly monotone.

We draw our second application from probability theory. To emphasize this, we denote the measure space by $(\Omega, \mathcal{F}, \mathbb{P})$ and we speak of random variables and events instead of measurable functions and measurable sets respectively. We note that if $k = 2$, then Theorem 3.1 says that correlated variables have nonnegative covariance, that is, $\mathbb{E}[f_1 f_2] - \mathbb{E}[f_1]\mathbb{E}[f_2] \geq 0$ (where $\mathbb{E}[f] := \int_\Omega f \mathrm{d}\mathbb{P}$ is the usual expectation).

We call the (real) random variables $\{X_0, X_1, \ldots, X_k\}$ *independent* if and only if, for every family of Borel sets $\{A_0, A_1, \ldots, A_k\}$, we have $\mathbb{P}(\cap_{i=0}^k \{X_i \in A_i\}) = \prod_{i=0}^k \mathbb{P}(X_i \in A_i)$, where $\mathbb{P}(X_i \in A_i)$ is shorthand for $\mathbb{P}(\{\omega \in \Omega : X_i(\omega) \in A_i\})$.

In order to make a specific example, let us think of the variable X_i ($i = 1, \dots, k$) as the (random) time made by the i-th contestant in an individual time trial bicycle race and let X_0 be our own (random) time; we suppose that each contestant is unaware of the results of the others (this is the independence hypothesis). If we know the probability of winning a one-to-one race against each of our competitors we may be interested, for instance, in estimating the probability of winning the race. Such estimates are possible as a consequence of Theorem 3.1; indeed, we have that

$$\mathbb{P}(\cap_{i=1}^k \{X_i \geq X_0\}) \geq \prod_{i=1}^k \mathbb{P}(X_i \geq X_0),$$

$$\mathbb{P}(\cap_{i=1}^k \{X_i \leq X_0\}) \geq \prod_{i=1}^k \mathbb{P}(X_i \leq X_0).$$

Thus the events $\{\{X_i \geq X_0\}\}_{i=1}^k$ (resp. $\{\{X_i \leq X_0\}\}_{i=1}^k$) are positively correlated (roughly speaking this means that knowing that $\{X_1 \geq X_0\}$ makes, for instance, the event $\{X_2 \geq X_0\}$ more likely than before).

The proof of these inequalities is straightforward. If we define $\mu(A) := \mathbb{P}(X_0 \in A)$ for all Borel sets $A \subseteq \mathbb{R}$, then, according to Fubini's Theorem,

$$\mathbb{P}(X_i \geq X_0) = \int_{\mathbb{R}} \mathbb{P}(X_i \geq t) \mathrm{d}\mu(t), \quad \mathbb{P}(\cap_{i=1}^k \{X_i \geq X_0\}) = \int_{\mathbb{R}} \prod_{i=1}^k \mathbb{P}(X_i \geq t) \mathrm{d}\mu(t),$$

$$\mathbb{P}(X_i \leq X_0) = \int_{\mathbb{R}} \mathbb{P}(X_i \leq t) \mathrm{d}\mu(t), \quad \mathbb{P}(\cap_{i=1}^k \{X_i \leq X_0\}) = \int_{\mathbb{R}} \prod_{i=1}^k \mathbb{P}(X_i \leq t) \mathrm{d}\mu(t).$$

Indeed,

$$\mathbb{P}(X_i \geq X_0) = \int_{\{(s,t) \in \mathbb{R}^2 : s \geq t\}} \mathrm{d}\nu(s) \mathrm{d}\mu(t) = \int_{\mathbb{R}} \int_{[t, +\infty)} \mathrm{d}\nu(s) \mathrm{d}\mu(t) = \int_{\mathbb{R}} \mathbb{P}(X_i \geq t) \mathrm{d}\mu(t),$$

where $\nu(A) := \mathbb{P}(X_i \in A)$ for all borel sets $A \subseteq \mathbb{R}$ and the first equality holds since X_i and X_0 are independent. The remaining cases are analogous. Note that $\{\mathbb{P}(X_i \geq t)\}_{i=1}^k$ and $\{\mathbb{P}(X_i \leq t)\}_{i=1}^k$ are both families of monotone (thus correlated) functions; Theorem 3.1 yields the claim. This example can be easily extended to a more interesting case: namely, when $\{X_1, \dots, X_k\}$ have identical laws and are independently conditioned to X_0 (see Chapters 4 and 6 of [1] for details). In this case one can prove that

$$\mathbb{P}(\cap_{i=1}^k \{X_i \in A\}) \geq \prod_{i=1}^k \mathbb{P}(X_i \in A), \qquad \forall A \subseteq \mathbb{R} \text{ Borel set.}$$

The proof makes use of Theorem 3.1 in its full generality but this example exceeds the purpose of this paper.

REFERENCES

[1] P. BILLINGSLEY, *Probability and Measure*, Wiley Series in Probability and Mathematical Statistics, John Wiley & Sons, New York, 1995.

[2] R.A. BRUALDI, Mathematical Notes: Comments and Complements, *Amer. Math. Monthly,* **84**(10) (1977), 803–807.

[3] I.S. GRADSHTEYN AND I.M. RYZHIK, *Tables of Integrals, Series, and Products*, 6th ed., Academic Press, San Diego, 2000.

[4] S.K. STEIN, An inequality in two monotonic functions, *Amer. Math. Monthly*, **83**(6) (1976), 469–471.