

Rearrangement Invariant, Coherent Risk Measures on L^0

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Abstract

By this paper, we give an answer to the problem of definition of coherent risk measures on rearrangement invariant, solid subspaces of L^0 with respect to some atom less probability space $(\Omega, \mathcal{F}, \mathbb{P})$. This problem was posed by F. Delbaen, while in this paper we proposed a solution via ideals of L^0 and the class of the dominated variation distributions, as well.

Keywords

Rearrangement Invariance, Dominated Variation, Moment-Index

1. Introduction

In (Delbaen, 2009), the problem of defining a risk measure on a *solid, rearrangement invariant* subspace of $L^0(\Omega, \mathcal{F}, \mathbb{P})$ -space of random variables with respect to some atomless probability space $(\Omega, \mathcal{F}, \mathbb{P})$. We recall that a vector space E , being a vector subspace of L^0 is called *rearrangement invariant* if for random variables $y, x \in L^0$, which have the same distribution, $x \in E$ implies $y \in E$. Also, the space E is *solid* if for random variables $y, x \in L^0$, $|y| \leq |x|, x \in E$, implies $y \in E$. In (Delbaen, 2009), there is an extensive treatment of this problem, related to the role of the spaces L^∞ and L^1 , compared to E , especially in (Delbaen, 2009). On the other hand, the whole paper (Delbaen, 2002) is devoted to the difficulties of defining coherent risk measures on subspaces of L^0 , while it is proved that if the probability space is atomless, no coherent risk measure is defined all over L^0 (Delbaen, 2002). Of course these attempts of moving from L^∞ to appropriately defined subspaces of L^0 , are related to the tail properties of the random variables in actuarial science and finance and more specifically to heavy-tailed distributed random variables. The actual problem behind these seminal article by F. Delbaen is since we cannot define a coherent risk measure on the entire L^0 , whether subspaces of L^0 which

are both alike L^∞ and preserve nice distributional properties (from the aspect of heavy-tails). Especially, we treat the rearrangement invariance in the sense of remaining in the same *class of distributions* and not by requiring distributional invariance. This is the topic of our paper.

2. Ideals of L^0 and Heavy-Tailed Distributions

It is well-known that since $L^0(\Omega, \mathcal{F}, \mathbb{P})$ is a Riesz space, being ordered by the pointwise- \mathbb{P} -a.e. partial ordering \geq , it would be taken as a Riesz subspace of \mathbb{R}^Ω . Hence, it may be considered to be an order-complete Riesz space. Let us take an element y of L^0 , which corresponds to a *heavy-tailed* random variable. This indicates that either for $y^+ = y \vee 0$ for $y^- = (-y) \vee 0$,

$$\mathbb{E}\left(e^{\varepsilon \cdot y^*}\right) = \infty,$$

for any real number $\varepsilon > 0$, where $* = +$ or $* = -$. Heavy-tailed random variables may not have even a finite moment $\mathbb{E}(|y|)$. On the other hand, according to (Aliprantis & Border, 1999), the principal ideal E_y generated by y in E , endowed by the norm

$$\|z\|_\infty = \inf \{ \lambda > 0 \mid |z| \leq \lambda |y| \},$$

is an AM-space with *order unit* $|y|$. We also have to mention the following relevant.

Lemma 2.1 *If $y > 0$, ($y \geq 0, y \neq 0$) and y is a heavy-tailed random variable, then every $z \in [-y, y]$ is a heavy-tailed random variable.*

Proof. Since $y \geq z, z \in [-y, y]$, we get that for the sets $y(t) = \{\omega \in \Omega \mid y(\omega) \leq t\}$, $z(t) = \{\omega \in \Omega \mid z(\omega) \leq t\}$, the inclusion $y(t) \subseteq z(t)$ holds, which implies $F_y(t) \leq F_z(t), t \in \mathbb{R}$ for the corresponding cumulative distribution functions. Since for the integral

$$\int_0^\infty e^{-\varepsilon t} dF_y(t) = \infty,$$

holds for any $\varepsilon > 0$, this implies

$$\int_0^\infty e^{-\varepsilon t} dF_z(t) = \infty,$$

for any $\varepsilon > 0$.

We recall the class \mathcal{D} of *dominated variation* distributions:

$$F \in \mathcal{D} \Leftrightarrow \limsup_{t \rightarrow +\infty} \frac{\bar{F}(tu)}{\bar{F}(t)} < \infty, \quad u \in (0, 1).$$

This class is a sub-class of heavy-tailed distributions, see (Cai & Tang, 2004).

Theorem 2.2 *If $F_y \in \mathcal{D}$, where \mathcal{D} denotes the class of dominated variation distributions respectively, then for every $z \in E_y$, $F_z \in \mathcal{D}$.*

Proof. According to what is proved in (Cai & Tang, 2004), the class \mathcal{D} is *convolution-closed*, namely if $F_x, F_y \in \mathcal{D}$, then $F_{x+y} \in \mathcal{D}$. First, we have to prove that if $F_y \in \mathcal{D}$, then $F_z \in \mathcal{D}$, for any $z \in E_y$. Since $z \in E_y$, there exists some $\lambda > 0$ such that $z \in [-\lambda \cdot y, \lambda \cdot y]$. But $F_{\lambda \cdot y} \in \mathcal{D}$. This is easy to prove, since if

$$\limsup_{t \rightarrow +\infty} \frac{\bar{F}_y(ut)}{\bar{F}_y(t)} < \infty,$$

for any $u \in (0, 1)$, then in order to prove that

$$\limsup_{t \rightarrow +\infty} \frac{\bar{F}_{\lambda \cdot y}(ut)}{\bar{F}_{\lambda \cdot y}(t)} < \infty,$$

for any $u \in (0, 1)$, then we get that the above limsup is equal to

$$\limsup_{r \rightarrow +\infty} \frac{\bar{F}_y(ru)}{\bar{F}_y(r)} < \infty,$$

for any $u \in (0,1)$. Hence, $F_{\lambda \cdot y} \in \mathcal{D}$. Moreover, we have to prove that if $z \in [-\lambda \cdot y, \lambda \cdot y]$, then $F_z \in \mathcal{D}$. From the previous Lemma,

$$\bar{F}_z(t) \leq \bar{F}_y(t), \quad \bar{F}_z(tu) \leq \bar{F}_y(tu), \quad u \in (0,1).$$

From the properties of the tail function of z we also have that since $ut < t$ for $u \in (0,1)$, then

$$\bar{F}_z(ut) \geq \bar{F}_z(t).$$

Hence,

$$1 \leq \frac{\bar{F}_z(ut)}{\bar{F}_z(t)} \leq \frac{\bar{F}_y(ut)}{\bar{F}_y(t)}.$$

Since $\frac{1}{\bar{F}_y(t)} \leq \frac{1}{\bar{F}_z(t)}$,

$$\limsup_{t \rightarrow +\infty} \frac{\bar{F}_z(tu)}{\bar{F}_z(t)} < \infty,$$

which is the desired conclusion.

Hence we obtain subspaces E of L^0 , which are actually the ideals E_y which satisfy the *rearrangement invariance* property, while they contain non-integrable distributions, in the sense that for any $z \in E_y$ there is a maximum p for which the moment $\mathbb{E}(|z|^p)$ exists in \mathbb{R} . Let us discuss more this question. A notion which is very important is the one of the *moment index*. We recall that the *moment index* for a non-negative random variable x is equal to

$$\mathbb{I}(x) = \sup \{v \mid \mathbb{E}(x^v) < \infty\}.$$

We also recall that if $F_x \in \mathcal{D}$, then $\mathbb{I}(x) < \infty$, see in (Seneta, 1976), (Tang & Tsitsiashvili, 2003). The use of the moment index in the specific case is that despite the validity of the (Delbaen, 2002), due to the fact that the elements of E_y distributions lie in the class \mathcal{D} , we assure that at least in the ideal E_y , we assure a general level of non-integrability of $z \in E_y$, given by a finite $\mathbb{I}(z)$. About the question whether the class \mathcal{D} is the greatest in which the specific Theorem holds, we have to mention that if we move up to the class of the *subexponential* distributions, it is not convolution-closed, see for example in (Leslie, 1989). As it is also well-known from (Aliprantis & Border, 1999), the dual space E_y^* of E_y is an AL-space, since the ideal E_y is an AM-space with unit $|y|$, as mentioned above. Hence, we keep the dual pair

$$\langle E_y, E_y^* \rangle,$$

for any of the y described above.

3. Expected-Shortfall on Ideals of L^0

Taking any $y > 0$ whose $F_y \in \mathcal{D}$, and defining the corresponding dual pair

$$\langle E_y, E_y^* \rangle,$$

we may define an Expected Shortfall-form risk measure on E_y . We have to notice that E_y satisfies both the order and the distributional rearrangement property, as a subspace of L^0 . This is due to the properties of the class \mathcal{D} of the dominated variation distributions. Hence we use (Kaina & Rüschendorf, 2009) of the dual (robust) representation of the usual Expected Shortfall in order to prove the following.

Theorem 3.1 *The functional $R: E_y \rightarrow \bar{\mathbb{R}}$, where*

$$R(x) = \sup_{f \in \left[0, \frac{1}{a}e\right]} f(-x)$$

is an $(E_{y,+}, y)$ -coherent risk measure, where $e \in E_{y,+} \setminus \{0\}$ is such that $\langle y, e \rangle = a, a > 0$.

Proof.

1)

$$R(x+t \cdot y) = \sup_{f \in \left[0, \frac{1}{a}e\right]} f(-x-t \cdot y) = \sup_{f \in \left[0, \frac{1}{a}e\right]} f(-x) - t,$$

for any $t \in \mathbb{R}$, due to the order completeness of the ideal E_y (y -Translation Invariance).

2)

$$R(x+z) = \sup_{f \in \left[0, \frac{1}{a}e\right]} f(-x-z) \leq \sup_{f \in \left[0, \frac{1}{a}e\right]} f(-x) + \sup_{f \in \left[0, \frac{1}{a}e\right]} f(-z) = R(x) + R(z),$$

for any $x, z \in E_y$ (Subadditivity).

3)

$$R(\lambda \cdot x) = \sup_{f \in \left[0, \frac{1}{a}e\right]} f(-\lambda \cdot x) = \lambda \cdot \sup_{f \in \left[0, \frac{1}{a}e\right]} f(-x),$$

for any $\lambda \in \mathbb{R}_+$ and $x \in E_y$ (Positive Homogeneity).

4) If $x \geq z$ then for any $f \in \left[0, \frac{1}{a}e\right]$ we get $f(-z) \geq f(-x)$. Hence by taking suprema all over

$f \in \left[0, \frac{1}{a}e\right]$, we get $R(z) \geq R(x)$ $x, z \in E_y$ ($E_{y,+}$ -Monotonicity).

Finally, if we suppose that the dual pair $\langle E_y^*, E_y \rangle$ is a symmetric Riesz pair, or else that E_y^* has order-continuous norm (see also (Aliprantis & Border, 1999)), then the values of R are finite since they represent the supremum value of a weak-star continuous linear functional on a weak-star compact set, which is the box of functionals $\left[0, \frac{1}{a}e\right]$. Otherwise, the infinity of the values of R may be excused by the presence of heavy-tailed distributions.

References

- Aliprantis, C. D., & Border, K. C. (1999). *Infinite Dimensional Analysis, A Hitchhiker's Guide* (2nd ed.). Springer. <http://dx.doi.org/10.1007/978-3-662-03961-8>
- Cai, J., & Tang, Q. (2004). On Max-Sum Equivalence and Convolution Closure of Heavy-Tailed Distributions and Their Applications. *Journal of Applied Probability*, 41, 117-130. <http://dx.doi.org/10.1239/jap/1077134672>
- Delbaen, F. (2002). Coherent Risk Measures on General Probability Spaces. *Advances in Finance and Stochastics: Essays in Honour of Dieter Sondermann*. Berlin: Springer-Verlag, 1-38. http://dx.doi.org/10.1007/978-3-662-04790-3_1
- Delbaen, F. (2009). Risk Measures for Non-Integrable Random Variables. *Mathematical Finance*, 19, 329-333. <http://dx.doi.org/10.1111/j.1467-9965.2009.00370.x>
- Kaina, M., & Rüschendorf, L. (2009). On Convex Risk Measures on L^1 -Spaces. *Mathematical Methods of Operations Research*, 69, 475-495. <http://dx.doi.org/10.1007/s00186-008-0248-3>
- Leslie, J. (1989). On the Non-Closure under Convolution of the Subexponential Family. *Journal of Applied Probability*, 26, 58-66. <http://dx.doi.org/10.2307/3214316>
- Seneta, E. (1976). *Regularly Varying Functions*. Lecture Notes in Mathematics, Vol. 508, Springer. <http://dx.doi.org/10.1007/BFb0079658>
- Tang, Q., & Tsitsiashvili, G. (2003). Precise Estimates for the Ruin Probability in Finite Horizon in a Discrete-Time Model with Heavy-Tailed Insurance and Financial Risks. *Stochastic Processes and Their Applications*, 108, 299-325. <http://dx.doi.org/10.1016/j.spa.2003.07.001>