

A NOTE ON A RECENT PAPER BY ZAKS,  
FROSTIG AND LEVIKSON

BY

KLAUS D. SCHMIDT

ABSTRACT

In the present paper we give a short proof of a result of Zaks, Frostig and Levikson [2006] on the solution of an optimization problem which is related to the problem of optimal pricing of a heterogeneous portfolio.

Following Zaks, Frostig and Levikson [2006], we consider a heterogeneous portfolio which is composed by  $k$  risk classes such that for each  $j \in \{1, \dots, k\}$  the risk class  $j$  contains  $n_j$  risks  $X_{j,1}, \dots, X_{j,n_j}$  which are assumed to be i.i.d. with finite first and second moments and non-zero variance. Then the total risk of risk class  $j$  is defined as

$$S_j := \sum_{i=1}^{n_j} X_{j,i}$$

Consider also  $r_1, \dots, r_k \in (0, \infty)$  and  $\alpha \in (0, 1)$ , and let  $z_{1-\alpha}$  denote the  $1-\alpha$  percentile of the standard normal distribution. The authors prove the following result:

**Theorem 1.** *The minimization problem*

*Minimize*

$$\sum_{j=1}^k \left( \frac{1}{r_j} E \left[ (S_j - n_j \pi_j)^2 \right] \right)$$

*over  $\pi_1, \dots, \pi_k$  subject to*

$$\sum_{j=1}^k n_j \pi_j = E \left[ \sum_{j=1}^k S_j \right] + z_{1-\alpha} \sqrt{\text{var} \left[ \sum_{j=1}^k S_j \right]}$$

*has a unique solution  $\pi_1^*, \dots, \pi_k^*$  and the identity*

$$\pi_j^* = \frac{1}{n_j} \left( E[S_j] + \frac{r_j}{\sum_{i=1}^k r_i} z_{1-\alpha} \sqrt{\text{var} \left[ \sum_{i=1}^k S_i \right]} \right)$$

holds for all  $j \in \{1, \dots, k\}$ .

Let now  $\mathbf{S}$  denote the random vector with coordinates  $S_1, \dots, S_k$  and let  $\mathbf{v} := E[\mathbf{S}]$ . Let also  $\mathbf{V}$  denote the diagonal matrix with diagonal elements  $r_1, \dots, r_k$ , let  $\mathbf{1}$  denote the vector with all coordinates being equal to one, and consider  $t \in \mathbb{R}$ . Using this notation, Theorem 1 can be stated in the following form, which suggests a simple proof based on the projection theorem in Hilbert spaces (see e.g. De Vylder [1996; Part III] or Swartz [1994; Section 6.6]):

**Theorem 1'.** *The minimization problem*

Minimize

$$E[(\mathbf{S} - \mathbf{p})' \mathbf{V}^{-1} (\mathbf{S} - \mathbf{p})]$$

over  $\mathbf{p}$  subject to  $\mathbf{1}' \mathbf{p} = \mathbf{1}' \mathbf{v} + t$

has a unique solution  $\mathbf{p}^*$  and the solution satisfies  $\mathbf{p}^* = \mathbf{v} + t(\mathbf{1}' \mathbf{V} \mathbf{1})^{-1} \mathbf{V} \mathbf{1}$ .

**Proof.** Since the matrix  $\mathbf{V}$  is symmetric and positive definite, the vector space  $L^2(\mathbb{R}^k)$  consisting of all  $k$ -dimensional random vectors having finite second moments is a Hilbert space under the inner product  $\langle \cdot, \cdot \rangle_{\mathbf{V}}$  given by

$$\langle \mathbf{X}, \mathbf{Y} \rangle_{\mathbf{V}} := E[\mathbf{X}' \mathbf{V}^{-1} \mathbf{Y}]$$

and the induced norm  $\|\cdot\|_{\mathbf{V}}$  given by

$$\|\mathbf{X}\|_{\mathbf{V}} := \langle \mathbf{X}, \mathbf{X} \rangle_{\mathbf{V}}^{1/2}$$

(Here, as usual, two random vectors  $\mathbf{X}, \mathbf{Y}$  are identified if  $P\{\mathbf{X} = \mathbf{Y}\} = 1$ .) Furthermore, the set

$$A := \{ \mathbf{p} \in \mathbb{R}^k \mid \mathbf{1}' \mathbf{p} = \mathbf{1}' \mathbf{v} + t \}$$

is a nonempty closed subset of  $L^2(\mathbb{R}^k)$ . Since  $A$  is convex, it follows from the projection theorem in Hilbert spaces that the minimization problem

Minimize

$$\|\mathbf{S} - \mathbf{p}\|_{\mathbf{V}}$$

over  $\mathbf{p} \in A$

has a unique solution  $\mathbf{p}^* \in A$ . Since  $A$  is even affine,  $\mathbf{p}^*$  is also the unique solution to the normal equations

$$\langle \mathbf{S} - \mathbf{p}^*, \mathbf{p} - \mathbf{p}^* \rangle_{\mathbf{V}} = 0$$

with  $\mathbf{p} \in A$  being arbitrary. Using the definition of the inner product  $\langle \cdot, \cdot \rangle_{\mathbf{V}}$ , the normal equations can also be written as

$$(\mathbf{v} - \mathbf{p}^*)' \mathbf{V}^{-1} (\mathbf{p} - \mathbf{p}^*) = 0$$

We now observe that every vector  $\mathbf{q}_\gamma := \mathbf{v} + \gamma \mathbf{V} \mathbf{1}$  with  $\gamma \in \mathbb{R}$  satisfies

$$(\mathbf{v} - \mathbf{q}_\gamma)' \mathbf{V}^{-1} (\mathbf{p} - \mathbf{q}_\gamma) = -\gamma (\mathbf{1}' \mathbf{p} - \mathbf{1}' \mathbf{q}_\gamma)$$

and that  $\mathbf{q}_\gamma \in A$  if and only if  $\gamma = t (\mathbf{1}' \mathbf{V} \mathbf{1})^{-1}$ . We have thus shown that the vector  $\mathbf{q} := \mathbf{v} + t (\mathbf{1}' \mathbf{V} \mathbf{1})^{-1} \mathbf{V} \mathbf{1}$  satisfies  $\mathbf{q} \in A$  and

$$(\mathbf{v} - \mathbf{q})' \mathbf{V}^{-1} (\mathbf{p} - \mathbf{q}) = 0$$

for all  $\mathbf{p} \in A$ . Therefore, we have  $\mathbf{q} = \mathbf{p}^*$ . □

#### REFERENCES

- DE VYLDER, E.F. (1996) *Advanced Risk Theory*. Bruxelles: Editions de l'Université de Bruxelles.  
 SWARTZ, C. (1994) *Measure, Integration and Function Spaces*. New Jersey – London: World Scientific.  
 ZAKS, Y., FROSTING, E., and LEVIKSON, B. (2006) Optimal pricing of a heterogeneous portfolio for a given risk level. *ASTIN Bulletin* **36**, 161-185.

KLAUS D. SCHMIDT  
*Lehrstuhl für Versicherungsmathematik*  
*Technische Universität Dresden*  
*D-01062 Dresden*  
*E-mail: klaus.d.schmidt@tu-dresden.de*