ON OPTIMAL PROPERTIES OF THE STOP LOSS REINSURANCE

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Let F(x) be the distribution function of the total amount ξ of claims of an insurance company. It is assumed that this company reduces its liability by means of one or several reinsurance arrangements. Let the remaining part of claims amount be equal to η , which is another random variable. Reinsurance arrangements are supposed to fulfil the following consistency condition: if $\xi = x$, then almost certainly $0 \le \eta \le x$. Presumably all reinsurance arrangements are supposed to fulfil the practice can be supposed to fulfil this requirement.

The problem is to find an optimal reinsurance arrangement, i.e. an optimal random variable η , in the sense that, the net reinsurance premium being given, the variance of η reaches its minimum. In other words, a variable η is looked for, which gives

 $E \{\eta\} = P = \text{const.}; V \{\eta\} = \text{minimum.}$

In the sequel we use conditional expectations in the sense defined by DOOB ([I]).

Let η be an arbitrary random variable satisfying the consistency condition. If

$$R(x) = E \{ \eta \mid \xi = x \},$$

then evidently $0 \leq R(x) \leq x$. Further we have

$$P = E \{\eta\} = \int_{\bullet}^{\bullet} E \{\eta \mid \xi = x\} dF = \int_{\bullet}^{\bullet} R(x) dF.$$

For the variance the inequality

$$V \{\eta\} + P^{2} = E \{\eta^{2}\} = \int_{0}^{\infty} E \{\eta^{2} | \xi = x\} dF \geq \int_{0}^{\infty} R^{2}(x) dF$$

holds true, since

$$\int_{0}^{\infty} E \{\eta^{2} | \xi = x\} dF - \int_{0}^{\infty} R^{2}(x) dF = \int_{0}^{\infty} E \{(\eta - R(x))^{2} | \xi = x\} dF \ge 0.$$

Clearly the arrangement $E \{ \eta | \xi \}$ gives also to the reinsurer a smaller variance than the original arrangement.

Hence we may restrict ourselves to variables $\eta = R(\xi)$, where R(x) is an arbitrary function (for which the above integrals exist) fulfilling the requirements

$$0 \leq R(x) \leq x; E \{ R (\xi) \} = P.$$

KAHN has proved ([2]), that a solution of the problem is then the Stop Loss treaty

$$R^*(x) = \begin{cases} x & \text{for } x < M \\ M & \text{for } x \ge M, \end{cases}$$

where M is chosen so that $E \{ R^* (\xi) \} = P$. In fact,

$$V \{ R (\xi) \} + P^{2} = \int_{0}^{\infty} R^{2} dF = \int_{0}^{\infty} (R - M)^{2} dF + 2MP - M^{2}$$

$$\geq \int_{0}^{M} (R - M)^{2} dF + 2MP - M^{2} \geq \int_{0}^{M} (x - M)^{2} dF + 2MP - M^{2}$$

$$= \int_{0}^{\infty} (R^{*} - M)^{2} dF + 2MP - M^{2} = V \{ R^{*} (\xi) \} + P^{2}.$$

REFERENCES

- [1] J. L. DOOB: Stochastic Processes. New York 1953.
- [2] P. M. KAHN: Some Remarks on a Recent Paper by Borch. The ASTIN Bulletin, Vol. I, Part V, pp. 265-272.