

SOME INEQUALITIES FOR STOP-LOSS PREMIUMS

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1. *A certain family of premium calculation principles*

In this paper any given risk S (a random variable) is assumed to have a (finite or infinite) mean. We enforce this by imposing $E[S^-] < \infty$.

Let then $v(t)$ be a twice differentiable function with

$$v'(t) > 0, v''(t) \geq 0, -\infty < t < +\infty$$

and let z be a constant with $0 \leq z \leq 1$.

We define the premium P as follows

$$P = \sup \{Q \mid -\infty < Q < +\infty, E[v(S - zQ)] > v((1 - z)Q)\} \quad (1)$$

or equivalently

$$P = \sup \{Q \mid -\infty < Q < +\infty, v^{-1} \circ E[v(S - zQ)] > (1 - z)Q\}. \quad (2)$$

Notation: $v^{-1}(\infty) = \infty$.

The definitions (1) and (equivalently) (2) are meaningful because of the

Lemma: a) $E[v(S - zQ)]$ exists for all $Q \in (-\infty, +\infty)$.

b) The set $\{Q \mid -\infty < Q < +\infty, E[v(S - zQ)] > v((1 - z)Q)\}$ is *not* empty.

Proof: a) $E[v^-(S - zQ)] \leq v^-(0) \cdot P[S \geq zQ] + v'(0) \int_{s < zQ} (zQ - S) dP(S)$
 $\leq v^-(0) \cdot P[S \geq zQ] + v'(0)[zQ + E(S^-)] < \infty$

b) Because of a) $E[v(S - zQ)]$ is always finite or equal to $+\infty$

If $v(-\infty) = -\infty$ then $E[v(S - zQ)] > v((1 - z)Q)$ is satisfied for sufficiently small Q . The left hand side of the inequality is a nonincreasing continuous function in P (strictly decreasing if $z > 0$), while the right hand

side is a nondecreasing continuous function in Q (strictly increasing if $z < 1$).

If $v(-\infty) = c$ finite then $E[v(S - zQ)] > c$

(otherwise S would need to be equal to $-\infty$ with probability 1) and again $E[v(S - zQ)] > v((1 - z)Q)$ is satisfied for sufficiently small Q .

From the lemma we conclude the following useful

Corrolary: There are two cases to be distinguished

a) *finite case:* There exists Q^* (finite) with

$$E[v(S - zQ^*)] = v((1 - z)Q^*) \quad (1^*)$$

or equivalently

$$v^{-1} \circ E[v(S - zQ^*)] = (1 - z)Q^* \quad (2^*)$$

then $P = Q^*$.

b) *infinite case:* Otherwise $P = +\infty$.

Proof: From the proof of the lemma it is obvious that Q^* under a) coincides with the supremum defining P .

Our premium calculation principle is determined by the choice of the function v and the constant z satisfying the above conditions. It satisfies the following very desirable postulates: For any risk S , for which the premium P exists,

$P_1 : P \geq E[S]$ $P_2 : P \leq \text{Max}[S]$
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Here $\text{Max}[S]$ denotes the right hand end point of the range of S .

Proof: For P_1 we start with equation (2) and make use of Jensen's inequality: P is the least upper bound of the set of Q 's for which

$$(1 - z)Q < v^{-1} \circ E[v(S - zQ)].$$

By Jensen's inequality

$$v^{-1} \circ E[v(S - zQ)] \geq v^{-1} \circ v(E[S - zQ]) = E[S] - zQ.$$

The set of Q 's for which

$Q < E[S]$ is hence a subset and its supremum $E[S]$ can not exceed the supremum P of the bigger set.

For P_2 we start with equation (2*) (only the case $\text{Max}[S] < \infty$ needs to be proved) and get

$$\begin{aligned} (1-z)P &= v^{-1} \circ E[v(S-zP)] \\ &\leq v^{-1} \circ \text{Max}[v(S-zP)] \\ &= v^{-1} \circ v(\text{Max}[S-zP]) \\ &= \text{Max}[S] - zP \end{aligned} \quad \text{q.e.d.}$$

Remarks:

1) If $z = 1$, we obtain the *principle of zero utility*,

$$P = \sup \{Q \mid E[u(Q-S)] < u(0)\}$$

by setting $u(t) = -v(-t)$.

2) If $z = 0$, we obtain the *mean value principle*,

$$P = v^{-1} \circ E[v(S)].$$

3) In the case where the function v is linear or exponential, the premium calculation principle does not depend on the value of z .

2. Partial Ordering among risks

Let $G(x)$, $H(x)$ be any distributions on the real line. Then we say that $G < H$, if

$$(PO) \int_t^{\infty} (x-t) dG(x) \leq \int_t^{\infty} (x-t) dH(x), \quad -\infty < t < \infty.$$

Condition (b) simply means that for any retention limit t the *net stoploss premium* for a risk whose cdf is G is not higher than the one for a risk whose cdf is H . We do allow the case where the integrals become infinite. Integration by parts leads to the following equivalent condition:

$$(PO') \int_0^{\infty} [1-G(x)] dx \leq \int_0^{\infty} [1-H(x)] dx.$$

The equivalence of (PO) and (PO') in the case of infinite integrals is e.g. proved in Feller II, page 150.

Let us now consider two stop-loss arrangements based on risks with cdf G and H , respectively. Let P_α^G, P_α^H denote the corresponding stop-loss premiums ($\alpha =$ retention limit). For example, P_α^H is obtained as the least upper bound of the set of Q 's for which

$$v((I - z)Q) < v(-zQ) H(\alpha) + \int_\alpha^\infty v(t - \alpha - zQ) dH(t) \quad (3)$$

and in the finite case as the unique solution of

$$v[(I - z)P_\alpha^H] = v(-zP_\alpha^H) H(\alpha) + \int_\alpha^\infty v(t - \alpha - zP_\alpha^H) dH(t) \quad (3^*)$$

The importance of the partial ordering introduced in this section becomes evident in the following theorem:

Theorem 1: Suppose $G < H$
Then $P_\alpha^G \leq P_\alpha^H, -\infty < \alpha < +\infty$

Proof: If $P_\alpha^H = \infty$ nothing is to be proved. We therefore assume P_α^H finite which implies $\int_0^\infty [I - H(x)] dx < \infty$ for all $t \in (-\infty, +\infty)$.

If we integrate in equation (3*) twice by parts, we obtain:

$$\begin{aligned} v((I - z)P_\alpha^H) &= v(-zP_\alpha^H) + \int_\alpha^\infty v'(t - \alpha - zP_\alpha^H) [I - H(t)] dt \\ &= v(-zP_\alpha^H) + v'(-zP_\alpha^H) \int_\alpha^\infty [I - H(t)] dt \\ &\quad + \int_\alpha^\infty v''(t - \alpha - zP_\alpha^H) \int_0^t [I - H(x)] dx dt. \end{aligned}$$

Now we estimate the last two terms from below, replacing H by G and using condition (PO'). By reversing the last step (integration by parts) we arrive at

$$v[(I - z)P_\alpha^H] \geq v(-zP_\alpha^H) + \int_\alpha^\infty v'(t - \alpha - zP_\alpha^H) [I - G(t)] dt$$

and therefore $P_\alpha^G \leq P_\alpha^H$ q.e.d.

We postpone examples to sections 3 and 4 and conclude this section with some useful lemmas. Their content is essentially that

the partial ordering is preserved under mixing and under convolution.

Lemma 1: Let $(G_n), (H_n)$ be sequences of distributions, and let (p_n) be a discrete probability distribution. If $G_n < H_n$ for all n , then

$$\sum_n p_n G_n < \sum_n p_n H_n.$$

Proof: Apply monotone convergence theorem

Lemma 2: If $G < H$, then

$$G * F < H * F.$$

Proof: To establish the validity of condition (PO') , we observe that

$$\begin{aligned} & \int_0^{\infty} [I - G * F(x)] dx \\ &= \int_0^{\infty} \int_{-\infty}^{\infty} [I - G(x - s)] dF(s) dx \end{aligned}$$

and by Fubini's theorem

$$= \int_{-\infty}^{\infty} \int_0^{\infty} [I - G(y)] dy dF(s).$$

The last expression shows that we obtain an upper bound if we replace G by H . q.e.d.

Lemma 3: If $G_i < H_i, (i = 1, 2, \dots, n)$, then

$$G_1 * G_2 * \dots * G_n < H_1 * H_2 * \dots * H_n.$$

Proof: Repeated application of Lemma 2 leads to

$$\begin{aligned} & G_1 * G_2 * G_3 * \dots * G_n \\ &< H_1 * G_2 * G_3 * \dots * G_n \\ &< H_1 * H_2 * G_3 * \dots * G_n \\ &< H_1 * H_2 * H_3 * \dots * G_n \quad \text{etc.} \end{aligned}$$

q.e.d.

3. Application 1: Dangerous Distributions

Definition: A distribution H is called *more dangerous* than a distribution G if (A) the first moments say μ_G, μ_H exist and $\mu_G \leq \mu_H$ and if (B) there is a constant β such that

$$\begin{aligned} G(x) &\leq H(x) \text{ for } x < \beta \\ G(x) &\geq H(x) \text{ for } x \geq \beta. \end{aligned}$$

Example 1: Let G be unimodal with $G(a-) = 0, G(b) = 1$ for $-\infty < a < b < \infty$. Let c, d be numbers such that $c \leq a, b \leq d$ and $(c + d)/2 \geq \mu_G$. Then the uniform distribution over the interval (c, d) is more dangerous than G .

Example 2: Let F be a distribution with $F(a-) = 0, F(b) = 1$ for $-\infty < a < b < \infty$. Let

$$G(x) = \begin{cases} 0 & \text{for } x < \mu_F \\ 1 & \text{for } x \geq \mu_F \end{cases}$$

and

$$H(x) = \begin{cases} 0 & \text{for } x < a \\ \frac{b - \mu_F}{b - a} & \text{for } a \leq x < b \\ 1 & \text{for } x \geq b. \end{cases}$$

Then F is more dangerous than G , and H is more dangerous than F .

Theorem 2: If H is more dangerous than G , then $G < H$.

Proof: Condition (PO') is obviously satisfied if $t \geq \beta$. If $t < \beta$, its validity can be seen as follows:

$$\begin{aligned} &\int_t^{\infty} [1 - G(x)] dx - \int_t^{\infty} [1 - H(x)] dx \\ &= \int_t^{\infty} [H(x) - G(x)] dx \\ &\leq \int_{-\infty}^{\infty} [H(x) - G(x)] dx = \mu_G - \mu_H \leq 0. \quad \text{q.e.d.} \end{aligned}$$

Illustration 1: Let $S = S_1 + S_2 + \dots + S_n$ be a sum of n independent risks. If we replace each of these risks by a more dangerous risk, the stop-loss premium for the sum of these new risks will be at least as high as the stop-loss premium for S (use Theorems 1, 2 and Lemma 3).

Illustration 2: Let S be a risk with a compound Poisson distribution, say with Poisson parameter λ and amount distribution $F(x)$. We assume that $F(0) = 0$ (only positive claims) and that $F(M) = 1$ for some $M > 0$ (a claim amount is at most M), and let μ denote the mean of F (i.e. the average claim amount). We compare S with the two compound Poisson risks S^μ, S^M with fixed claim amounts μ, M , respectively, and Poisson parameters $\lambda, \Lambda = \lambda(\mu/M)$, respectively. (Observe that $E(S^\mu) = E(S) = E(S^M)$.) From Example 2 (with $a = 0, b = M$), Lemmas 1, 3, and Theorems 1, 2 we obtain inequalities for the corresponding stop-loss premiums:

$$P_x^\mu \leq P_x \leq P_x^M.$$

In the case of net stop-loss premiums the second inequality has been proved by Gagliardi and Straub (Mitteilungen Vereinigung schweizerischer Versicherungsmathematiker, 1974, Heft 2).

4. *Application 2: Random sums of positive risks*

In this section we shall compare a distribution of the form

$$G = (1 - q) F^{*0} + qF, \quad 0 \leq q \leq 1 \tag{4}$$

with one of the more general form

$$H = \sum_{n=0}^{\infty} p_n F^{*n} \tag{5}$$

where

$$0 \leq p_n \leq 1, \quad \sum_{n=0}^{\infty} p_n = 1.$$

Theorem 3: Suppose $F(0) = 0$
 If $\sum_{n=1}^{\infty} np_n = q$, then $G < H$, where G, H are given by (4), (5).

Proof: Firstly, we show that

$$F < \frac{n-1}{n} F^{*0} + \frac{1}{n} F^{*n}, \quad n = 1, 2, \dots \quad (6)$$

which is a special case of Theorem 3.

To show the validity of condition (PO) we introduce the independent random variables X_1, X_2, \dots, X_n with common distribution F . Then condition (PO) is equivalent to

$$\sum_{i=1}^n E[(X_i - t)_+] \leq (n-1)(-t)_+ + E[(\sum_{i=1}^n X_i - t)_+].$$

But the corresponding inequality is satisfied for any outcomes of X_1, X_2, \dots, X_n .

Secondly, we show that $G < H$ in the general case. Since

$$H = \sum_{n=1}^{\infty} n p_n \left[\frac{n-1}{n} F^{*0} + \frac{1}{n} F^{*n} \right] + (1-q) F^{*0}$$

$$G = \sum_{n=1}^{\infty} n p_n F + (1-q) F^{*0}$$

this follows from equation (6) and Lemma 1.

Illustration: Individual versus collective model: The individual model is described by n numbers q_i , $0 < q_i \leq 1$, and n distributions F_i with $F_i(0) = 0$. We have in mind a portfolio consisting of n components. Then q_i is the probability that a claim occurs in component i , and F_i is the distribution of its amount. Let

$$S^{\text{ind}} = S_1 + S_2 + \dots + S_n$$

denote the total claims of the portfolio, where

$$\text{Prob}(S_i = 0) = 1 - q_i$$

$$\text{Prob}(S_i \leq x) = 1 - q_i + q_i F_i(x), \quad x > 0$$

for $i = 1, 2, \dots, n$. We assume that S_1, S_2, \dots, S_n are independent and denote the stop-loss premium for S^{ind} by P_{α}^{ind} ($\alpha =$ retention limit).

A collective model is assigned to the individual model in a well known fashion: Let S^{coll} denote the compound Poisson random variable with

$$\text{Poisson parameter } \lambda = \sum_{i=1}^n q_i$$

$$\text{Amount distribution } F = \sum_{i=1}^n q_i/\lambda F_i.$$

Let P_α^{coll} denote the stop-loss premium for S^{coll} . By applying Theorem 3 to each of the n components (replacing S_i by a compound Poisson random variable with Poisson parameter q_i and amount distribution F_i), we recognize from Theorem 1 and Lemma 3 that $P_\alpha^{\text{ind}} \leq P_\alpha^{\text{coll}}$. Thus a cautious reinsurer will prefer the collective model to the individual model.