

## Explicit Asymptotics for the Ruin Probability with Risky Investment Included\*

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### Abstract

In this paper, we investigate the ruin probability of a discrete-time risk model, in which the surplus of an insurance business is currently invested into a risky asset. Using a purely probabilistic treatment, we establish explicit asymptotic relations for the infinite-time ruin probabilities, hence we extend a recent result of Tang and Tsitsiashvili (2003) to the infinite-time case.

**Keywords:** Asymptotics, regular variation, ruin probability, stochastic equation.

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### § 1. Introduction

Following Nyrhinen (1999), Tang and Tsitsiashvili (2003), and Chen and Xie (2005), we consider a discrete-time risk model, in which the surplus of the insurance company is currently invested into a risky asset which may lead to a negative return in each year. Denote by  $A_n \in (-\infty, \infty)$  the net income (the total incoming premium minus the total claim amount) within year  $n$  and by  $r_n \in (-1, \infty)$  the return rate at year  $n$ ,  $n = 1, 2, \dots$ . Let the initial surplus be  $x \geq 0$ . Hence, if we assume that the net income  $A_n$  is calculated at the beginning of year  $n$ , then the surplus accumulated till the end of year  $n$ , characterized by  $S_n$ , satisfies the recurrence equation

$$S_0 = x \geq 0, \quad S_n = (1 + r_n)(S_{n-1} + A_n), \quad n = 1, 2, \dots; \quad (1.1)$$

alternatively, if we assume that the net income  $A_n$  is calculated at the end of year  $n$ , then the surplus accumulated till the end of year  $n$ , characterized by  $T_n$ , satisfies the recurrence equation

$$T_0 = x \geq 0, \quad T_n = (1 + r_n)T_{n-1} + A_n, \quad n = 1, 2, \dots. \quad (1.2)$$

Throughout the paper, we assume that the net incomes  $A_n$ ,  $n = 1, 2, \dots$ , constitute a sequence of independent, identically distributed (i.i.d.) random variables (r.v.'s), that the return rates  $r_n$ ,  $n = 1, 2, \dots$ , also constitute a sequence of i.i.d. r.v.'s, and that the two sequences  $\{A_n, n = 1, 2, \dots\}$  and  $\{r_n, n = 1, 2, \dots\}$  are independent.

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Write

$$X_n = -A_n, \quad Y_n = \frac{1}{1+r_n}, \quad n = 1, 2, \dots \quad (1.3)$$

The r.v.  $X_n$  is the net payout during year  $n$  and the r.v.  $Y_n$  is the discount factor from year  $n$  to year  $n-1$ ,  $n = 1, 2, \dots$ . In what follows, we write by  $X$  and  $Y$  the generic r.v.'s of the sequences  $\{X_n, n = 1, 2, \dots\}$  and  $\{Y_n, n = 1, 2, \dots\}$  and write by  $F$  and  $G$  the distribution functions (d.f.'s) of the r.v.'s  $X$  and  $Y$ , respectively. Clearly, the r.v.  $Y$  is strictly positive. We assume that the tail probability  $\bar{F}(x) = 1 - F(x) = P(X > x)$  satisfies  $\bar{F}(x) > 0$  for any real number  $x$ . In the terminology of Norberg (1999) and Tang and Tsitsiashvili (2003), we call  $X$  the insurance risk and  $Y$  the financial risk.

Corresponding to the surplus processes (1.1) and (1.2), we define the ruin probabilities within finite time  $n = 1, 2, \dots$  and infinite time as

$$\psi_S(x, n) = P\left(\min_{0 \leq k \leq n} S_k < 0 | S_0 = x\right), \quad \psi_T(x, n) = P\left(\min_{0 \leq k \leq n} T_k < 0 | T_0 = x\right), \quad (1.4)$$

respectively,

$$\psi_S(x) = P\left(\min_{0 \leq k < \infty} S_k < 0 | S_0 = x\right), \quad \psi_T(x) = P\left(\min_{0 \leq k < \infty} T_k < 0 | T_0 = x\right). \quad (1.5)$$

In this paper we are interested in the asymptotic behavior of these probabilities.

Hereafter, all limit relationships are for  $x \rightarrow \infty$  unless stated otherwise; for two positive functions  $a(x)$  and  $b(x)$ , we write  $a(x) \lesssim b(x)$  if  $\limsup a(x)/b(x) \leq 1$ , write  $a(x) \gtrsim b(x)$  if  $\liminf a(x)/b(x) \geq 1$ , and write  $a(x) \sim b(x)$  if both.

Nyrhinen (1999) investigated the asymptotic behavior of the ruin probability  $\psi_S(x)$ . In terms of the model described above, we can obtain a combination of Theorems 3.3 and 3.4 of Nyrhinen (1999) as follows: if (1)  $w = \sup\{t | EY^t \leq 1\} \in (0, \infty)$ , (2)  $EY^t$  and  $E|X|^t$  are finite for some  $t > w$ , (3)  $\bar{F}(0) > 0$ , and (4) some of the convolution powers of the distribution of  $\log Y$  has a non-trivial absolutely continuous component, then the relation

$$\psi_S(x) \sim Cx^{-w} \quad (1.6)$$

holds for some positive, but implicit, constant  $C$ .

The objective of the present paper is to establish explicit asymptotic relations for the ruin probabilities  $\psi_S(x)$  and  $\psi_T(x)$  under some other assumptions on the tails of the r.v.'s  $X$  and  $Y$ . In the proof, we will use a recent result of Tang and Tsitsiashvili (2003), who investigated the finite-time ruin probabilities  $\psi_S(x, n)$  and  $\psi_T(x, n)$  under the assumptions that  $F$  is heavy tailed and that the tail  $\bar{G}$  is dominated by the tail  $\bar{F}$ . We will also apply some results obtained by Vervaat (1979) in the literature of stochastic difference equations.

The rest of this paper consists of three sections: Section 2 presents the main results after recalling an important class of heavy-tailed distributions, Section 3 collects some lemmas, and Section 4 proves the theorems.

## § 2. Main Results

We will assume that the d.f.  $F$  of the insurance risk  $X$  has a regularly varying tail, denoted by  $F \in \mathcal{R}$ . By definition, a d.f.  $F$  concentrated on  $(-\infty, \infty)$  belongs to the class  $\mathcal{R}$  if there is some  $\alpha \geq 0$  such that

$$\lim_{x \rightarrow \infty} \frac{\overline{F}(xy)}{\overline{F}(x)} = y^{-\alpha} \tag{2.1}$$

for any  $y > 0$ . For simplicity, we denote by  $F \in \mathcal{R}_{-\alpha}$  the regularity property in (2.1). In this case we have

$$\overline{F}(x) = x^{-\alpha} c(x) \exp \left\{ \int_a^x \frac{\varepsilon(y)}{y} dy \right\}, \quad x > a, \tag{2.2}$$

for some  $a > 0$ , where  $c(x) \rightarrow c \in (0, \infty)$  and  $\varepsilon(x) \rightarrow 0$ ; see Bingham et al. (1987, page 21). The corresponding r.v.  $X$  satisfies

$$\mathbf{E}(X^+)^p < \infty \text{ for } 0 \leq p < \alpha, \quad \mathbf{E}(X^+)^p = \infty \text{ for } p > \alpha, \tag{2.3}$$

where  $x^+ = \max\{x, 0\}$ .

Under the assumptions that  $F$  belongs to a heavy-tailed distribution class, which is slightly larger than the class  $\mathcal{R}$ , and that the tail  $\overline{G}$  is dominated by the tail  $\overline{F}$  (to be precise,  $\mathbf{E}Y^p < \infty$  for some  $p$  larger than the upper Matuszewska index of the d.f.  $F$ ), Tang and Tsitsiashvili (2003, Theorem 5.1 and Remark 5.1) obtained precise asymptotic estimates for the finite-time ruin probabilities  $\psi_S(x, n)$  and  $\psi_T(x, n)$  for each fixed  $n = 1, 2, \dots$ . The really applicable results of their paper were obtained in the case that

$$F \in \mathcal{R}_{-\alpha} \quad \text{and} \quad \mathbf{E}Y^p < \infty \quad \text{for some } p > \alpha > 0. \tag{2.4}$$

In this case, it holds for each fixed  $n = 1, 2, \dots$  that

$$\psi_S(x, n) \sim \frac{1 - (\mathbf{E}Y^\alpha)^n}{1 - \mathbf{E}Y^\alpha} \overline{F}(x); \tag{2.5}$$

see Tang and Tsitsiashvili (2003, Theorem 5.2(3) and Remark 5.1).

If further we assume that

$$\mathbf{E}Y^\alpha < 1, \tag{2.6}$$

then the term  $(\mathbf{E}Y^\alpha)^n$  in relation (2.5) vanishes as  $n$  increases. Though relation (2.5) was only proved by Tang and Tsitsiashvili (2003) to hold for fixed  $n$ , we intuitively believe that the infinite-time ruin probability  $\psi_S(x)$  should satisfy

$$\psi_S(x) \sim \frac{1}{1 - \mathbf{E}Y^\alpha} \overline{F}(x). \tag{2.7}$$

The following result proves that this is true.

**Theorem 2.1** Under assumptions (2.4) and (2.6), the asymptotic relation (2.7) holds.

Theorem 2.1 successfully establishes an asymptotic relation for the ruin probability  $\psi_S(x)$  in a fully explicit form.

As for the ruin probability  $\psi_T(x)$ , we have a similar explicit asymptotic relation below.

**Theorem 2.2** Under assumptions (2.4) and (2.6), it holds that

$$\psi_T(x) \sim \frac{\mathbf{E}Y^\alpha}{1 - \mathbf{E}Y^\alpha} \bar{F}(x). \quad (2.8)$$

### § 3. Some Lemmas

We first point out a simple relationship between the ruin probabilities indexed by  $S$  and those indexed by  $T$ .

**Lemma 3.1** For the risk model introduced in Section 1, the ruin probabilities defined by (1.4) and (1.5) satisfy the relations

$$\psi_T(x, n) = \int_0^\infty \psi_S(x/y, n) G(dy), \quad n = 1, 2, \dots, \quad (3.1)$$

and

$$\psi_T(x) = \int_0^\infty \psi_S(x/y) G(dy). \quad (3.2)$$

**Proof** Iterating the recurrence equation (1.1) yields that

$$S_0 = x, \quad S_n = x \prod_{j=1}^n (1 + r_j) + \sum_{i=1}^n A_i \prod_{j=i}^n (1 + r_j), \quad n = 1, 2, \dots. \quad (3.3)$$

We write the discounted value of the surplus  $S_n$  in (3.3) as

$$\tilde{S}_0 = x, \quad \tilde{S}_n = S_n \prod_{j=1}^n Y_j = x + \sum_{i=1}^n A_i \prod_{j=i}^{n-1} Y_j = x - \sum_{i=1}^n X_i \prod_{j=1}^{i-1} Y_j,$$

where  $\prod_{j=1}^0 = 1$  by convention. It is clear that for each  $n = 1, 2, \dots$ ,

$$\psi_S(x, n) = \mathbf{P}\left(\max_{1 \leq k \leq n} \sum_{i=1}^k X_i \prod_{j=1}^{i-1} Y_j > x\right), \quad \psi_S(x) = \mathbf{P}\left(\max_{1 \leq k < \infty} \sum_{i=1}^k X_i \prod_{j=1}^{i-1} Y_j > x\right). \quad (3.4)$$

Similarly, it holds that

$$\psi_T(x, n) = \mathbf{P}\left(\max_{1 \leq k \leq n} \sum_{i=1}^k X_i \prod_{j=1}^i Y_j > x\right), \quad \psi_T(x) = \mathbf{P}\left(\max_{1 \leq k < \infty} \sum_{i=1}^k X_i \prod_{j=1}^i Y_j > x\right). \quad (3.5)$$

Hence by the i.i.d. assumptions made in Section 1, relations (3.1) and (3.2) holds. #

The lemma below is a combination of several results of Vervaat (1979).

**Lemma 3.2** Let  $\{(\tilde{X}_n, \tilde{Y}_n), n = 1, 2, \dots\}$  be a sequence of i.i.d. random pairs with generic random pair  $(\tilde{X}, \tilde{Y})$ . Consider the stochastic difference equation

$$\tilde{V}_n = \tilde{Y}_n \tilde{V}_{n-1} + \tilde{X}_n, \quad n = 1, 2, \dots. \quad (3.6)$$

If  $-\infty \leq \mathbf{E} \log \tilde{Y} < 0$  and  $\mathbf{E}(\log |\tilde{X}|)^+ < \infty$ , then the r.v.'s  $\tilde{V}_n$  converges in distribution to some real-valued r.v.  $V_\infty$ , which is invariant in distribution for all initial r.v.'s  $\tilde{V}_0$ .

**Proof** By Theorem 1.6(b, c) of Vervaat (1979), the stochastic equation

$$\tilde{V} =^d \tilde{Y}\tilde{V} + \tilde{X}, \quad \tilde{V} \text{ is independent of } (\tilde{X}, \tilde{Y}), \tag{3.7}$$

has a solution, where  $=^d$  denotes equality in distribution. Then, by Theorem 1.5(i) of Vervaat (1979), this solution is unique in distribution and  $\tilde{V}_n$  defined by (3.6) converges in distribution to some  $V_\infty(\tilde{V}_0)$ , say, for any initial r.v.  $\tilde{V}_0$ . Since, by Lemma 1.1 of Vervaat (1979), any limit r.v.  $V_\infty(\tilde{V}_0)$  should be a solution of equation (3.7), we prove that for all  $\tilde{V}_0$  the r.v.'s  $V_\infty(\tilde{V}_0)$  are identical in distribution. #

The following lemma is well known and is from Proposition of Feller (1971, p. 278) or Lemma 1.3.1 of Embrechts et al. (1997).

**Lemma 3.3** Let  $F_1$  and  $F_2$  be two d.f.'s concentrated on  $[0, \infty)$ . If  $F_1 \in \mathcal{R}_{-\alpha}$  and  $F_2 \in \mathcal{R}_{-\alpha}$  for some  $\alpha \geq 0$ , then their convolution  $F_1 * F_2 \in \mathcal{R}_{-\alpha}$  and  $\overline{F_1 * F_2}(x) \sim \overline{F_1}(x) + \overline{F_2}(x)$ .

The following lemma is from Breiman (1965).

**Lemma 3.4** Let  $X$  and  $Y$  be two independent r.v.'s distributed by  $F$  and  $G$ , respectively, where  $Y$  is nonnegative. If d.f.  $F \in \mathcal{R}_{-\alpha}$  for some  $0 < \alpha < \infty$  and  $\mathbf{E}Y^p < \infty$  for some  $p > \alpha$ , then

$$\lim_{x \rightarrow \infty} \frac{\mathbf{P}(XY > x)}{\mathbf{P}(X > x)} = \mathbf{E}Y^\alpha.$$

## § 4. Proofs of the Main Results

### 4.1 Proof of Theorem 2.1

From (3.4) and (2.5), it is clear that for each  $n = 1, 2, \dots$ ,

$$\psi_S(x) \geq \psi_S(x, n) \sim \frac{1 - (\mathbf{E}Y^\alpha)^n}{1 - \mathbf{E}Y^\alpha} \overline{F}(x).$$

It follows that for each  $n = 1, 2, \dots$ ,

$$\liminf_{x \rightarrow \infty} \frac{\psi_S(x)}{\overline{F}(x)} \geq \frac{1 - (\mathbf{E}Y^\alpha)^n}{1 - \mathbf{E}Y^\alpha}.$$

Letting  $n \rightarrow \infty$  on the right-hand side,

$$\psi_S(x) \gtrsim \frac{1}{1 - \mathbf{E}Y^\alpha} \overline{F}(x). \tag{4.1}$$

It remains to derive a corresponding asymptotic upper bound for the ruin probability  $\psi_S(x)$ . To this end, from (3.4) we derive that

$$\psi_S(x) \leq \mathbf{P}\left(\sum_{i=1}^{\infty} X_i^+ \prod_{j=1}^{i-1} Y_j > x\right). \tag{4.2}$$

For each  $n = 1, 2, \dots, \infty$ , set

$$U_n = \sum_{i=1}^n X_i^+ \prod_{j=1}^{i-1} Y_j.$$

By the i.i.d. assumptions, it is easy to see that for each  $n = 1, 2, \dots$ ,

$$U_n = {}^d \sum_{i=1}^n X_i^+ \prod_{j=i+1}^n Y_j = V_n, \quad (4.3)$$

where  $\prod_{j=n+1}^n Y_j$  is equal to 1 by convention. Clearly, with  $V_0 = 0$ , the sequence  $\{V_n, n = 1, 2, \dots\}$  satisfies the recurrence equation

$$V_n = Y_n V_{n-1} + X_n^+ \quad \text{for } n = 1, 2, \dots. \quad (4.4)$$

Now we check the convergence in distribution of the sequence  $\{V_n, n = 1, 2, \dots\}$ . By assumption (2.6) we easily understand that  $-\infty \leq \mathbf{E} \log Y < 0$  should hold since the function  $f(t) = \mathbf{E} Y^t$  is convex in  $t \in [0, \alpha]$  and  $f'_+(0) = \mathbf{E} \log Y$ . Hence by Lemma 3.2,  $V_n$  converges in distribution to a r.v.  $V_\infty$ , say, which is invariant for all  $V_0$ . In view of (4.3), this actually proves that  $U_\infty$  is finite almost surely and is equal to  $V_\infty$  in distribution. Specifically, we choose  $V_\infty$  to be independent of the sequences  $\{X_n, n = 1, 2, \dots\}$  and  $\{Y_n, n = 1, 2, \dots\}$ .

Knowing  $F \in \mathcal{R}_{-\alpha}$ , we announce that

$$\mathbf{P}(U_\infty > x) = \mathbf{P}(V_\infty > x) \leq \mathbf{P}(V_0 > x), \quad x \geq 0, \quad (4.5)$$

for some nonnegative initial r.v.  $V_0$  with a d.f. from the class  $\mathcal{R}_{-\alpha}$ . For this purpose, we choose a nonnegative r.v.  $Z$  independent of the sequences  $\{X_n, n = 1, 2, \dots\}$  and  $\{Y_n, n = 1, 2, \dots\}$  such that  $\mathbf{P}(Z > x) \sim c\bar{F}(x)$  for some positive constant  $c$ , which will be specified later. By Lemma 3.4 we know that the d.f. of the r.v.  $Y_1 Z$  belongs to the class  $\mathcal{R}_{-\alpha}$  and  $\mathbf{P}(Y_1 Z > x) \sim c\mathbf{E} Y^\alpha \bar{F}(x)$ ; by this and Lemma 3.3 we further know that  $\mathbf{P}(Y_1 Z + X_1^+ > x) \sim (c\mathbf{E} Y^\alpha + 1)\bar{F}(x)$ . Hence, if we choose  $c > 0$  sufficiently large such that  $c\mathbf{E} Y^\alpha + 1 < c$ , then there is some constant  $x_0 > 0$  such that for all  $x > x_0$ ,  $\mathbf{P}(Y_1 Z + X_1^+ > x) \leq \mathbf{P}(Z > x)$ . By this inequality we can prove that for all  $x \geq 0$ ,

$$\mathbf{P}(Y_1 Z + X_1^+ > x | Z > x_0) \leq \mathbf{P}(Z > x | Z > x_0). \quad (4.6)$$

In fact, for  $0 \leq x \leq x_0$ , inequality (4.6) trivially holds since  $\mathbf{P}(Z > x | Z > x_0) = 1$ ; for  $x > x_0$ , inequality (4.6) can be verified in the following way:

$$\begin{aligned} \mathbf{P}(Y_1 Z + X_1^+ > x | Z > x_0) &= \frac{\mathbf{P}(Y_1 Z + X_1^+ > x, Z > x_0)}{\mathbf{P}(Z > x_0)} \leq \frac{\mathbf{P}(Y_1 Z + X_1^+ > x)}{\mathbf{P}(Z > x_0)} \\ &\leq \frac{\mathbf{P}(Z > x)}{\mathbf{P}(Z > x_0)} = \mathbf{P}(Z > x | Z > x_0). \end{aligned}$$

We identify the initial r.v.  $V_0$  such that it is independent of the sequences  $\{X_n, n = 1, 2, \dots\}$  and  $\{Y_n, n = 1, 2, \dots\}$  and is equal in distribution to the r.v.  $Z$  conditional on  $(Z > x_0)$ . Hence,

$$\mathbf{P}(V_0 > x) = \mathbf{P}(Z > x | Z > x_0) \sim \frac{c}{\mathbf{P}(Z > x_0)} \bar{F}(x). \quad (4.7)$$

It follows from (4.6) that for all  $x \geq 0$ ,

$$\mathbf{P}(V_1 > x) = \mathbf{P}(Y_1 V_0 + X_1^+ > x) = \mathbf{P}(Y_1 Z + X_1^+ > x | Z > x_0) \leq \mathbf{P}(V_0 > x).$$

Successively,

$$\begin{aligned} P(V_2 > x) &= P(Y_2 V_1 + X_2^+ > x) \leq P(Y_2 V_0 + X_2^+ > x) = P(Y_1 V_0 + X_1^+ > x) \\ &= P(V_1 > x) \leq P(V_0 > x). \end{aligned}$$

Applying the mathematical induction method we know that for all  $n = 1, 2, \dots$  and all  $x \geq 0$ ,

$$P(V_n > x) \leq \dots \leq P(V_2 > x) \leq P(V_1 > x) \leq P(V_0 > x).$$

Since  $V_n$  converges in distribution to  $V_\infty$ , taking  $n \rightarrow \infty$  yields that for all  $x \geq 0$ ,

$$P(U_\infty > x) = P(V_\infty > x) \leq P(V_0 > x),$$

as announced in (4.5).

We continue the proof of Theorem 2.1. For any  $\varepsilon \in (0, 1)$  and any  $n = 1, 2, \dots$ , we split the probability on the right-hand side of inequality (4.2) into two parts as

$$\begin{aligned} \psi_S(x) &\leq P\left(\sum_{i=1}^n X_i^+ \prod_{j=1}^{i-1} Y_j > (1 - \varepsilon)x\right) + P\left(\sum_{i=n+1}^\infty X_i^+ \prod_{j=1}^{i-1} Y_j > \varepsilon x\right) \\ &= I_1(x, \varepsilon, n) + I_2(x, \varepsilon, n). \end{aligned}$$

Clearly, from (2.5) and (3.4) with  $X_i$  being replaced by  $X_i^+$ , we obtain that

$$I_1(x, \varepsilon, n) \sim \frac{1 - (\mathbf{E}Y^\alpha)^n}{1 - \mathbf{E}Y^\alpha} \bar{F}((1 - \varepsilon)x).$$

Since  $V_\infty$  is independent of the sequences  $\{X_n, n = 1, 2, \dots\}$  and  $\{Y_n, n = 1, 2, \dots\}$ ,

$$\begin{aligned} I_2(x, \varepsilon, n) &= P\left(\left(\sum_{i=n+1}^\infty X_i^+ \prod_{j=n+1}^{i-1} Y_j\right) \prod_{j=1}^n Y_j > \varepsilon x\right) = P\left(V_\infty \prod_{j=1}^n Y_j > \varepsilon x\right) \\ &\leq P\left(V_0 \prod_{j=1}^n Y_j > \varepsilon x\right) \sim \frac{c(\mathbf{E}Y^\alpha)^n}{P(Z > x_0)} \bar{F}(\varepsilon x), \end{aligned}$$

where in the last step we applied relation (4.7) and Lemma 3.4. Thus, for any  $\varepsilon \in (0, 1)$  and any  $n = 1, 2, \dots$ ,

$$\psi_S(x) \lesssim \frac{1 - (\mathbf{E}Y^\alpha)^n}{1 - \mathbf{E}Y^\alpha} \bar{F}((1 - \varepsilon)x) + \frac{c(\mathbf{E}Y^\alpha)^n}{P(Z > x_0)} \bar{F}(\varepsilon x).$$

It follows from  $F \in \mathcal{R}_{-\alpha}$  that

$$\begin{aligned} \limsup_{x \rightarrow \infty} \frac{\psi_S(x)}{\bar{F}(x)} &\leq \frac{1 - (\mathbf{E}Y^\alpha)^n}{1 - \mathbf{E}Y^\alpha} \limsup_{x \rightarrow \infty} \frac{\bar{F}((1 - \varepsilon)x)}{\bar{F}(x)} + \frac{c(\mathbf{E}Y^\alpha)^n}{P(Z > x_0)} \limsup_{x \rightarrow \infty} \frac{\bar{F}(\varepsilon x)}{\bar{F}(x)} \\ &= \frac{1 - (\mathbf{E}Y^\alpha)^n}{1 - \mathbf{E}Y^\alpha} (1 - \varepsilon)^{-\alpha} + \frac{c(\mathbf{E}Y^\alpha)^n}{P(Z > x_0)} \varepsilon^{-\alpha}. \end{aligned}$$

Since  $n$  and  $\varepsilon$  are arbitrary and  $\mathbf{E}Y^\alpha < 1$ , first letting  $n \rightarrow \infty$  and then letting  $\varepsilon \rightarrow 0$  lead to the desired result that

$$\psi_S(x) \lesssim \frac{1}{1 - \mathbf{E}Y^\alpha} \bar{F}(x). \tag{4.8}$$

Combining (4.1) and (4.8) we obtain (2.7). This ends the proof. #

#### 4.2 Proof of Theorem 2.2

Introduce a survival d.f.  $R_S$  by  $R_S(x) = (1 - \psi_S(x))1_{(x \geq 0)}$ , which is a standard d.f. concentrated on  $[0, \infty)$  with a mass  $R_S(\{0\}) = 1 - \psi_S(0)$  at 0. Theorem 2.1 has proved that  $R_S \in \mathcal{R}_{-\alpha}$ . Hence, applying Lemma 3.4 to relation (3.2) yields that

$$\psi_T(x) \sim \text{E}Y^\alpha \psi_S(x) \sim \frac{\text{E}Y^\alpha}{1 - \text{E}Y^\alpha} \bar{F}(x).$$

This ends the proof. #

#### References

- [1] Bingham, N.H., Goldie, C.M., Teugels, J.L., *Regular Variation*, Cambridge University Press, Cambridge, 1987.
- [2] Breiman, L., On some limit theorems similar to the arc-sin law, *Theor. Probability Appl.*, **10**(1965), 323–331.
- [3] Chen, Y., Xie, X., The finite time ruin probability with the same heavy-tailed insurance and financial risks, *Acta Math. Appl. Sin. Engl. Ser.*, **21**(1)(2005), 153–156.
- [4] Embrechts, P., Klüppelberg, C., Mikosch, T., *Modelling Extremal Events for Insurance and Finance*, Springer-Verlag, Berlin, 1997.
- [5] Feller, W., *An Introduction to Probability Theory and Its Applications*, Vol. II, Second edition John Wiley & Sons, Inc., New York-London-Sydney, 1971.
- [6] Norberg, R., Ruin problems with assets and liabilities of diffusion type, *Stochastic Process. Appl.*, **81**(2)(1999), 255–269.
- [7] Nyrhinen, H., On the ruin probabilities in a general economic environment, *Stochastic Process. Appl.*, **83**(2)(1999), 319–330.
- [8] Tang, Q., Tsitsiashvili, G., Precise estimates for the ruin probability in finite horizon in a discrete-time model with heavy-tailed insurance and financial risks, *Stochastic Process. Appl.*, **108**(2)(2003), 299–325.
- [9] Vervaat, W., On a stochastic difference equation and a representation of nonnegative infinitely divisible random variables, *Adv. in Appl. Probab.*, **11**(4)(1979), 750–783.

## 在风险投资下破产概率的一个渐近显式

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在本文中, 我们研究了一个离散时间风险模型的破产概率. 在此风险模型中, 保险公司的剩余资本被用于进行风险投资. 我们运用纯概率的手法建立了无限时间破产概率的渐近显式, 从而将 Tang 和 Tsitsiashvili (2003) 近期的一个结果推广到了无限时间的场合.

**关键词:** 渐近式, 正则变化, 破产概率, 随机方程.

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