

Ruin Probability of Risk Model with Interference under Stochastic Interest Rate

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Abstract—In this paper, we discuss the ruin probability of risk model with interference term under stochastic interest rate on the basis of classical risk model. And the properties of surplus process are discussed. The expression of ruin probability is given by using the properties of surplus process. The ruin probability considering interest rate and inflation is obtained.

Keywords—Surplus process; Random interest rate; Wiener process; Interference; ruin probability.

I. INTRODUCTION

In the general risk theoretical model, the interest rate is set as a constant, but many economic behaviors are long-term. During this period, government policies, economic cycles and other factors will cause uncertainty, that is, interest rate volatility, which means the randomness of interest rate. Moreover, it is recognized that the risk generated by the randomness of interest rates (for insurance companies) may be quite large and cannot be dispersed by increasing the sales volume of policies. However, the risks generated by the randomness of accidents (for insurance companies) can be dispersed by selling a large number of policies. In this sense, interest rate risk should be paid more attention than the risk of accidents. So this paper studies the ruin probability model under stochastic interest rate. Moreover, because insurance companies have uncertain income and payment in their daily operations, random interference terms are added to the model.

The surplus process of the insurance company is

$$R(t) = (x + ct)(1 + I) - S(t) + \sigma W(t)$$

Among :

$R(t)$: Surplus of the insurance company at the moment t ;

$x = R(0)$: The initial surplus of the insurance company;

c : Premium collected in unit time;

I : Random interest rate and $E(I) = i$, $Var(I) = \theta^2$

;

$S(t)$: Prime Minister's compensation amount up to the moment t , and

$$S(t) = \sum_{k=1}^{N(t)} X_k.$$

Among : X_1, X_2, Λ independent of each other and distributed in the same distribution function $F(\cdot)$;

$\{N(t), t \geq 0\}$ is the

Poisson of the parameter μ , and X_1, X_2, Λ is independent of each other.

$W(t)$: Standard Wiener process, represents the uncertain earnings and payments of the insurance company. $\sigma > 0$.

Arder

$$T = \inf\{t | R(t) < 0\}$$

indicates the moment of bankruptcy, and

$$\psi(x) = P(T < \infty)$$

is probability of bankruptcy.

In this paper, we discuss the properties of surplus process, and give the expression of ruin probability by using the properties of surplus process.

II. NATURE OF SURPLUS PROCESS

Nature 1 Surplus process $\{R(t), t \geq 0\}$ is stable independent increment.

Prove $t_0 < t_1 < \Lambda < t_n$, We can get

$$R(t_i) - R(t_{i-1}) =$$

$$c(1 + I)(t_i - t_{i-1}) - [S(t_i) - S(t_{i-1})] + \sigma[W(t_i) - W(t_{i-1})].$$

So

$$1 + I, S(t_1) - S(t_0), S(t_2) - S(t_1), \Lambda,$$

$$S(t_n) - S(t_{n-1}), W(t_1) - W(t_0), W(t_2) - W(t_1), \Lambda,$$

$$W(t_n) - W(t_{n-1}) \text{ are mutually independent. Surplus}$$

process $\{R(t), t \geq 0\}$ has independent increments.

Because

$$R(t+s) - R(t) = cs(1 + I) - [S(t+s) - S(t)] + \sigma[W(t+s) - W(t)]$$

$$, cs(1 + I), S(t+s) - S(t), W(t+s) - W(t) \text{ have the}$$

same distribution respectively for every

$$t \geq 0. \text{ So, } R(t+s) - R(t)$$

has the same distribution for every $t \geq 0$, $\{R(t), t \geq 0\}$ has

smooth increments.

So, surplus process $\{R(t), t \geq 0\}$ has smooth increments.

Nature2 $E[R(t)] = (x + ct)(1 + i) - \mu p_1 t$,

$$\text{Var}[R(t)] = (x + ct)^2 \theta^2 + \mu p_2 t + \sigma^2 t.$$

Among $p_k = \int x^k dF(x)$, $k = 1, 2$.

Prove For t

$$\begin{aligned} E[R(t)] &= E[(x + ct)(1 + I) - S(t) + \sigma W(t)] \\ &= (x + ct)[1 + E(I)] - E[S(t)] + \sigma E[W(t)] \end{aligned}$$

Because

$$E(I) = i,$$

$$E[W(t)] = 0,$$

$$E[S(t)] = E\{E[S(t)|N(t)]\} = E[p_1 N(t)] = p_1 E[N(t)] = \mu p_1 t.$$

So

$$E[R(t)] = (x + ct)(1 + i) - \mu p_1 t.$$

Because of the random process I , $\{S(t), t \geq 0\}$, $\{W(t), t \geq 0\}$ are mutually independent, so

$$\text{Var}[R(t)] = (x + ct)^2 \text{Var}(1 + I) + \text{Var}[S(t)] + \sigma^2 \text{Var}[W(t)].$$

$$\text{Var}[W(t)] = t,$$

$$\text{Var}(1 + I) = \theta^2,$$

$$\text{Var}[S(t)] = E\{\text{Var}[S(t)|N(t)]\} + \text{Var}\{E[S(t)|N(t)]\} = \mu p_2 t,$$

So

$$\text{Var}[R(t)] = (x + ct)^2 \theta^2 + \mu p_2 t + \sigma^2 t.$$

In order to ensure the stable operation of the insurance company, the insurance premium income should be greater than the claim amount in unit time. According to nature 2, it is necessary to assume that $c(1 + i) > \mu p_1$.

III. RUIN PROBABILITY

Definition $M_x(r)$ is the moment generating function of individual claim amount. R is nonzero positive solution of the equation

$$-rc(1 + i) + \mu[M_x(r) - 1] + \frac{1}{2}\sigma^2 r^2 = 0 \quad (1)$$

R is called adjustment coefficient, the equation (1) is called adjustment coefficient equation.

lemma The solution of the adjustment coefficient equation (1).

Prove Let

$$g(r) = -rc(1 + i) + \mu[M_x(r) - 1] + \frac{1}{2}\sigma^2 r^2 = 0, \text{ 则}$$

$$\frac{dg(r)}{dr} = -c(1 + i) + \mu E(Xe^{rx}) + \sigma^2 r,$$

$$\frac{d^2 g(r)}{dr^2} = \mu E(X^2 e^{rx}) + \sigma^2 > 0.$$

So $g(r)$ is a lower convex function. The equation (1) is at most two solutions, and $g(0) = 0$, $r = 0$ is a trivial solution. Because

$$\left. \frac{dg(r)}{dr} \right|_{r=0} = -c(1 + i) + \mu p_1 < 0.$$

And $r \rightarrow +\infty$, $g(r) \rightarrow +\infty$, So the positive solution R is unique.

The following theorem gives the general expression of ruin probability.

$$\textbf{Theorem1} \quad \psi(x) = \frac{E[e^{-Rx(1+I)}]}{E[e^{-RR(T)+RcT(I-i)}] | T < \infty}.$$

R is the adjustment coefficient.

Prove $t > 0$, $r > 0$, we get

$$E[e^{-rR(t)}] = E[e^{-rR(t)} | T \leq t] P(T \leq t) + E[e^{-rR(t)} | T > t] P(T > t) \quad (2)$$

Because $R(t) = (x + ct)(1 + I) - S(t) + \sigma W(t)$,

$$\begin{aligned} E[e^{-rR(t)}] &= E[\exp[-rx(1 + I) - rct(1 + I) + rS(t) - r\sigma W(t)]] \\ &= E\{\exp[-rx(1 + I) - rct(1 + I) + \mu[M_x(r) - 1] + \frac{1}{2}\sigma^2 r^2 t]\}. \end{aligned} \quad (3)$$

Because

$$\begin{aligned} R(t) &= R(T) + [R(t) - R(T)] = \\ &= R(T) + c(1 + I)(t - T) - [S(t) - S(T)] + \sigma[W(t) - W(T)]. \end{aligned}$$

For T , $S(t) - S(T)$, $W(t) - W(T)$, $R(T)$ and I are mutually independent. $S(t) - S(T)$ is the compound Poisson distribution of $\mu(t - T)$. $W(t) - W(T) \sim N(0, t - T)$.

So

$$\begin{aligned} E[e^{-rR(t)} | T \leq t] P(T \leq t) &= \\ E\{\exp[-rR(T) - rc(1 + I)(t - T) + \mu(t - T)(M_x(r) - 1) \\ + \frac{1}{2}\sigma^2 r^2 t] | T \leq t\} P(T \leq t). \end{aligned} \quad (4)$$

So

$$-Rc(1 + i) + \mu[M_x(R) - 1] + \frac{1}{2}\sigma^2 R^2 = 0.$$

(3) become

$$E[e^{-rR(t)}] = E\{\exp[-Rx(1 + I) - Rct(I - i)]\}.$$

(4) become

$$\begin{aligned} E[e^{-rR(t)} | T \leq t] P(T \leq t) &= \\ E\{\exp[-RR(T) - Rc(t - T)(I - i)] | T \leq t\} P(T \leq t) &= \\ E\{\exp[-Rct(I - i)]\} E\{\exp[-RR(T) + RcT(I - i)] | T \leq t\} P(T \leq t) &= \\ E\{\exp[-RR(T) + RcT(I - i)] | T \leq t\} P(T \leq t) \rightarrow \\ E\{\exp[-RR(T) + RcT(I - i)] | T \leq \infty\} \psi(x) \quad (t \rightarrow \infty). \end{aligned}$$

For

$$E[e^{-rR(t)} | T > t] P(T > t). \quad (5)$$

Because

$$E[R(t)] = (x + ct)(1 + i) - \mu p_1 t,$$

$$\text{Var}[R(t)] = (x + ct)^2 \theta^2 + \mu p_2 t + \sigma^2 t.$$

Let $\alpha = c(1+i) - \mu p_1 > 0$, so

$$q(t) = x(1+i) + \alpha - [(x+ct)^2 \theta^2 + \mu p_2 t + \sigma^2 t]^{\frac{1}{3}}.$$

And let $t \rightarrow \infty$, we get $q(t) > 0$.

So, (5) become

$$\begin{aligned} E[e^{-RR(t)} | T > t, 0 \leq R(t) \leq q(t)] P(T > t, 0 \leq R(t) \leq q(t)) + \\ E[e^{-RR(t)} | T > t, R(t) > q(t)] P(T > t, R(t) > q(t)) \\ \leq P(R(t) \leq q(t)) + e^{-Rq(t)}. \end{aligned}$$

Applying Chebyshev inequality, we get

$$\begin{aligned} P(R(t) \leq q(t)) \\ = P(R(t) \leq x(1+i) + \alpha - [(x+ct)^2 \theta^2 + \mu p_2 t + \sigma^2 t]^{\frac{1}{3}}) \leq \\ P(|R(t) - (x(1+i) + \alpha)| \geq [(x+ct)^2 \theta^2 + \mu p_2 t + \sigma^2 t]^{\frac{1}{3}}) \leq \\ \frac{\text{Var}(R(t)) [(x+ct)^2 \theta^2 + \mu p_2 t + \sigma^2 t]^{\frac{2}{3}}}{[(x+ct)^2 \theta^2 + \mu p_2 t + \sigma^2 t]^{\frac{1}{3}}}. \end{aligned}$$

So

$$\begin{aligned} E[e^{-RR(t)}] = E\{\exp[-Rx(1+I) - Rct(I-i)]\} \leq \\ E\{\exp[-RR(T) - Rct(I-i)] | T < t\} P(T \leq t) \\ + [(x+ct)^2 \theta^2 + \mu p_2 t + \sigma^2 t]^{\frac{1}{3}} + e^{-Rq(t)}. \end{aligned}$$

So

$$\begin{aligned} E\{\exp[-Rx(1+I)]\} \leq E\{\exp[-RR(T) \\ + Rct(I-i)] | T < t\} P(T \leq t) + \\ \frac{[(x+ct)^2 \theta^2 + \mu p_2 t + \sigma^2 t]^{\frac{1}{3}} + e^{-Rq(t)}}{E[e^{-Rct(I-i)}]}. \end{aligned}$$

Let $t \rightarrow \infty$, we get $P(T \leq t) \rightarrow \psi(u)$, $e^{-Rq(t)} \rightarrow 0$, and

$$E[e^{-Rct(I-i)}] = \int_0^{\infty} e^{-Rct(x-i)} p(x) dx. \quad (6)$$

So (6) become

$$\begin{aligned} E[e^{-Rct(I-i)}] = \int_0^{\infty} e^{-Rct(x-i)} p(x) dx > \\ \int_{x_0-\delta}^{x_0+\delta} e^{-Rct(x-i)} p(x) dx > 2\delta e^{-Rct(x_0+\delta-i)} \rightarrow \infty (t \rightarrow \infty). \end{aligned}$$

Let $t \rightarrow \infty$, we get

$$\frac{[(x+ct)^2 \theta^2 + \mu p_2 t + \sigma^2 t]^{\frac{1}{3}} + e^{-Rq(t)}}{E[e^{-Rct(I-i)}]} \rightarrow 0.$$

and

$$E[e^{-Rx(1+I)}] = E[e^{-RR(T)+RcT(I-i)} | T < \infty] \psi(x).$$

即

$$\psi(x) = \frac{E[e^{-Rx(1+I)}]}{E[e^{-RR(T)+RcT(I-i)} | T < \infty]}.$$

Inference 1 If $I = 0$, we get

$$\psi(x) = \frac{e^{-Rx}}{E[e^{-RR(T)} | T < \infty]}.$$

It is the ruin probability of the model with interference risk without considering the interest rate factor. [6] discussed that Liver tumor division in restorative pictures has been generally considered as of late, of which the Level set models show an

uncommon potential with the advantage of overall optima and functional effectiveness. The Gaussian mixture model (GMM) and Expected Maximization for liver tumor division are introduced. [7] discussed that In surgical planning and cancer treatment, it is crucial to segment and measure a liver tumor's volume accurately. Because it would involve automation, standardisation, and the incorporation of complete volumetric information, accurate automatic liver tumor segmentation would substantially affect the processes for therapy planning and follow-up reporting.

Inference 2 If $I = i$, I is constant interest rate, we get

$$\psi(x) = \frac{E[e^{-Rx(1+i)}]}{E[e^{-RR(T)} | T < \infty]}.$$

That is the ruin probability of the risk model with interference under constant interest rate.

IV. MODEL CONSIDERING INTEREST RATE AND INFLATION

If J is the inflation rate, Surplus process is

$$R(t) = (x+ct)(1+I-J) - S(t) + \sigma W(t).$$

We get the theorem 2.

$$\textbf{Theorem 2} \quad \psi(x) = \frac{E[e^{-Rx(1+I-J)}]}{E\{e^{-RR(T)+RcT[(I-i)+(J-j)]} | T < \infty\}}.$$

R is the adjustment coefficient.

REFERENCES

- [1] Zhigang Xie, Tianxiong Han. Risk Theory and Life Insurance Actuaries [M], Tianjin: Nankai University Press, 2000, pp. 102-122
- [2] Minping Qian, Guanglu Gong. Stochastic Process Theory [M]. Beijing: Peking University Press. 1997, pp. 112-125
- [3] Yinghua Dong, Hanjun Zhang. Ruin probability of double Poisson risk model with interference [J]. Journal of Mathematical Theory and Application, 2003.23 (1): 98-101
- [4] Jinzhi Li, Kelin Qiao. Ruin Model of Random Factors [J]. Journal of Yunnan University, 2003, 25 (1): 9-12.
- [5] Waters H R. Probability of ruin for a risk process with claims cost inflation[J]. Scand Actuarial Journal, 1983, 2: 148-164.
- [6] Christo Ananth, M Kameswari, Densy John Vadakkan, Dr. Niha.K., "Enhancing Segmentation Approaches from Fuzzy-MPSO Based Liver Tumor Segmentation to Gaussian Mixture Model and Expected Maximization", Journal Of Algebraic Statistics, Volume 13, Issue 2, June 2022, pp. 788-797
- [7] Christo Ananth, S. Amutha, K. Niha, Djabbarov Botirjon Begimovich, "Enhancing Segmentation Approaches from Super Pixel Division Algorithm to Hidden Markov Random Fields with Expectation Maximization (HMR-EM)", International Journal of Early Childhood Special Education, Volume 14, Issue 05, 2022, pp. 2400-2410