

Computing Ruin Probability in Generalized Risk Processes under Constant Interest Force

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Abstract The aim of this paper is to build an exact formula for ruin probability of generalized risk processes under constant interest force with sequences of random variables such that these sequences are usually assumed to be positive integer – valued random variables. An exact formula for finite time ruin (non-ruin) probabilities are derive by using technique of classical probability. A numerical example is given to illustrate results.

Keywords Ruin Probability, Non –ruin Probability, Homogeneous Markov Chain

1. Introduction

Claude Lefèvre and Stéphane Loisel[1] studied the problem of ruin in the classical compound binomial and compound Poisson risk models. Their primary purpose is to extend those models which is an exact formula derived by Picard and Lefèvre[9] for the probability of (non-ruin) ruin within finite time. Hong N.T.T. (see [7]) recently built an exact formula for finite time ruin (non-ruin) probability for model:

$$U_t = u + \sum_{i=1}^t X_i - \sum_{i=1}^t Y_i$$

With u, t, X_i, Y_i are positive integer number.

However, Claude Lefèvre and Stéphane Loisel[1] did not provide an exact formula for ruin probability of generalized risk processes under constant interest force with sequences of random variables such that these sequences are usually assumed to be positive integer – valued random variables, with surplus process $\{U_t\}_{t \geq 1}$ written as

$$U_t = u(1+r)^t + \sum_{i=1}^t X_i(1+r)^{t-i+1} - \sum_{i=1}^t Y_i(1+r)^{t-i} \quad (1.1)$$

where $U_0 = u$ is initial surplus ($u > 0$), r is constant interest ($r > 0$), u and t are positive integer numbers,

$X = \{X_i\}_{i \geq 1}$ and $Y = \{Y_j\}_{j \geq 1}$ take values in a finite set of positive integer numbers; X and Y are assumed to be independent.

With these assumptions, the aim of this paper is to build an

exact formula for finite time ruin (non-ruin) probability of model (1.1). In our study, we extended the result of Hong N. T. T for model (1.1) with any $r > 0$. This is the first time that gives an exact formula for ruin (non-ruin) probability for model (1.1) whose exact formula for finite time ruin (non-ruin) probability are derived by using technique of classical probability.

The paper is organized as follows; in Section 2, we build an exact formula for ruin (non-ruin) probability for model

(1.1) with $X = \{X_i\}_{i \geq 1}$ and $Y = \{Y_j\}_{j \geq 1}$ being

independent and identically distributed positive integer – valued random variables, X and Y are assumed to be independent. An extended result in Section 2 with X and Y being homogeneous Markov chains is given in Section 3. A numerical example is give to illustrate these results in Section 4. Finally, we conclude our paper in Section 5.

2. Computing Ruin Probability of Generalized Risk Processes under Constant Interest Force with Sequences of Independent and Identically Distributed Random Variables

Let model (1.1). We assume that:

Assumption 2.1. u and t are positive integer numbers.

Assumption 2.2 $X = \{X_n\}_{n \geq 1}$ is a sequence of independent and identically distributed random variables, X_n take values in a finite set of positive integer numbers

$E_X = \{1, 2, \dots, M\}$ with $p_k = P(X_1 = k) (k \in E_X)$,

$$0 \leq p_k \leq 1, \sum_{k=1}^M p_k = 1.$$

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Assumption 2.3 $Y = \{Y_n\}_{n \geq 1}$ is a sequence of independent and identically distributed random variables, Y_n take values in a finite set of positive integer numbers $E_Y = \{1, 2, \dots, N\}$ with $q_k = P(Y_1 = k) (k \in E_Y)$,

$$0 \leq q_k \leq 1, \sum_{k=1}^N q_k = 1.$$

Assumption 2.4 X and Y are assumed to be independent.

From (1.1), we have:

$$U_t = u(1+r)^t + \sum_{i=1}^t X_i(1+r)^{t-i+1} - \sum_{i=1}^{t-1} Y_i(1+r)^{t-i} - Y_t. \quad (2.1)$$

Supposing that the ruin time is defined by $T_u = \inf \{j : U_j < 0\}$, where $\inf \emptyset = \infty$.

We define the finite time ruin (non-ruin) probabilities of model (1.1) with Assumption 2.1 to Assumption 2.4, respectively, by

$$\psi_t^{(1)}(u) = P(T_u \leq t) = P\left(\bigcup_{j=1}^t (U_j < 0)\right), \quad (2.2)$$

$$\begin{aligned} \varphi_t^{(1)}(u) &= 1 - \psi_t^{(1)}(u) \\ &= P(T_u \geq t+1) = P\left(\bigcap_{j=1}^t (U_j \geq 0)\right). \end{aligned} \quad (2.3)$$

Throughout this paper, we denote $A \stackrel{as}{=} B$ if

$$P((A \setminus B) \cup (B \setminus A)) = 0$$

To establish a formula for $\psi_t^{(1)}(u), \varphi_t^{(1)}(u)$, we first proof the following Lemma.

Lemma 2.1. Any u and $\{x_i\}_{i=1}^t, \{y_i\}_{i=1}^t$ are positive integer numbers.

With p being a positive integer number and $1 \leq p \leq t-1$ satisfies:

$$y_p \leq u(1+r)^p + \sum_{k=1}^p x_k(1+r)^{p+1-k} - \sum_{k=1}^{p-1} y_k(1+r)^{p-k}, \quad (2.4)$$

then

$$u(1+r)^{p+1} + \sum_{k=1}^{p+1} x_k(1+r)^{p+2-k} - \sum_{k=1}^p y_k(1+r)^{p+1-k} \geq 1. \quad (2.5)$$

Proof.

From (2.4), we have

$$\begin{aligned} y_p &\leq u(1+r)^p + \sum_{k=1}^p x_k(1+r)^{p+1-k} - \sum_{k=1}^{p-1} y_k(1+r)^{p-k} \\ &\Leftrightarrow -y_p \geq -u(1+r)^p - \sum_{k=1}^p x_k(1+r)^{p+1-k} + \sum_{k=1}^{p-1} y_k(1+r)^{p-k} \end{aligned}$$

Implies

$$\begin{aligned} &u(1+r)^{p+1} + \sum_{k=1}^{p+1} x_k(1+r)^{p+2-k} - \sum_{k=1}^p y_k(1+r)^{p+1-k} \\ &= u(1+r)^{p+1} + \sum_{k=1}^{p+1} x_k(1+r)^{p+2-k} - \sum_{k=1}^{p-1} y_k(1+r)^{p+1-k} - y_p(1+r) \\ &\geq u(1+r)^{p+1} + \sum_{k=1}^{p+1} x_k(1+r)^{p+2-k} - \sum_{k=1}^{p-1} y_k(1+r)^{p+1-k} \\ &\quad - \left(u(1+r)^p + \sum_{k=1}^p x_k(1+r)^{p+1-k} - \sum_{k=1}^{p-1} y_k(1+r)^{p-k} \right) (1+r) \\ &= x_{p+1}(1+r) \geq 1. \end{aligned}$$

Hence (2.5) holds.

This completes the proof.

Next, we give an exact formula for finite time non-ruin (ruin) probability of model (1.1).

Theorem 2.1. Let model (1.1) satisfy Assumption 2.1 to Assumption 2.4, then finite time non-ruin probability of model (1.1) is defined by

$$\begin{aligned} \varphi_t^{(1)}(u) &= \sum_{x_1, x_2, \dots, x_t=1}^M p_{x_1} p_{x_2} \dots p_{x_t} \\ &\quad \left(\sum_{1 \leq y_1 \leq g_1} \sum_{1 \leq y_2 \leq g_2} \dots \sum_{1 \leq y_t \leq g_t} q_{y_1} q_{y_2} \dots q_{y_t} \right), \end{aligned} \quad (2.6)$$

where

$$\begin{aligned} g_1 &= \min \{ [u(1+r) + x_1(1+r)], N \}, \\ g_2 &= \min \left\{ \left[u(1+r)^2 + \sum_{k=1}^2 x_k(1+r)^{3-k} - \sum_{k=1}^1 y_k(1+r)^{2-k} \right], N \right\}, \\ &\dots \\ g_t &= \min \left\{ \left[u(1+r)^t + \sum_{k=1}^t x_k(1+r)^{t+1-k} - \sum_{k=1}^{t-1} y_k(1+r)^{t-k} \right], N \right\}, \end{aligned}$$

In addition, $[u(1+r) + x_1(1+r)]$ is integer part of the $u(1+r) + x_1(1+r)$.

Proof.

Firstly, we have

$$\begin{aligned} A &:= \bigcap_{j=1}^t (U_j \geq 0) = (U_1 \geq 0) \cap (U_2 \geq 0) \cap \dots \cap (U_t \geq 0) \\ &= (Y_1 \leq u(1+r) + X_1(1+r)) \cap \\ &\quad \left(Y_2 \leq u(1+r)^2 + \sum_{k=1}^2 X_k(1+r)^{3-k} - \sum_{k=1}^1 Y_k(1+r)^{2-k} \right) \cap \\ &\quad \left(Y_3 \leq u(1+r)^3 + \sum_{k=1}^3 X_k(1+r)^{4-k} - \sum_{k=1}^2 Y_k(1+r)^{3-k} \right) \cap \\ &\quad \dots \cap \left(Y_t \leq u(1+r)^t + \sum_{k=1}^t X_k(1+r)^{t+1-k} - \sum_{k=1}^{t-1} Y_k(1+r)^{t-k} \right) \end{aligned} \quad (2.7)$$

By Assumption 2.2, we let $X_1 = x_1, X_2 = x_2, \dots, X_t = x_t$ satisfying: $1 \leq x_1, x_2, \dots, x_t \leq M$.
with x_1, x_2, \dots, x_t being positive integer numbers and

Let

$$A_{x_1 x_2 \dots x_t} = (X_1 = x_1) \cap (X_2 = x_2) \cap \dots \cap (X_t = x_t).$$

Since $X = \{X_n\}_{n \geq 1}$ is a sequence of independent random variables then

$$P(A_{x_1 x_2 \dots x_t}) = P[(X_1 = x_1) \cap (X_2 = x_2) \cap \dots \cap (X_t = x_t)] = P(X_1 = x_1) \cdot P(X_2 = x_2) \dots P(X_t = x_t) = p_{x_1} p_{x_2} \dots p_{x_t} \quad (2.8)$$

Hence, (2.7) is given

$$A \stackrel{as}{=} \bigcup_{x_1, x_2, \dots, x_t=1}^M \left(\{(X_1 = x_1) \cap (X_2 = x_2) \cap \dots \cap (X_t = x_t)\} \cap B_{x_1 x_2 \dots x_t} \right) = \bigcup_{x_1, x_2, \dots, x_t=1}^M (A_{x_1 x_2 \dots x_t} \cap B_{x_1 x_2 \dots x_t}), \quad (2.9)$$

where

$$\begin{aligned} B_{x_1 x_2 \dots x_t} \stackrel{as}{=} & (Y_1 \leq u(1+r) + x_1(1+r)) \cap \left(Y_2 \leq u(1+r)^2 + \sum_{k=1}^2 x_k(1+r)^{3-k} - \sum_{k=1}^1 Y_k(1+r)^{2-k} \right) \cap \\ & \left(Y_3 \leq u(1+r)^3 + \sum_{k=1}^3 x_k(1+r)^{4-k} - \sum_{k=1}^2 Y_k(1+r)^{3-k} \right) \cap \dots \\ & \cap \left(Y_t \leq u(1+r)^t + \sum_{k=1}^t x_k(1+r)^{t+1-k} - \sum_{k=1}^{t-1} Y_k(1+r)^{t-k} \right) \end{aligned} \quad (2.10)$$

By Assumption 2.3, we let $Y_1 = y_1, Y_2 = y_2, \dots, Y_{t-1} = y_{t-1}$ with y_1, y_2, \dots, y_{t-1} being positive integer numbers and satisfying: $1 \leq y_1, y_2, \dots, y_t \leq N$.

Let

$$\begin{aligned} g_1 &= \min \{ [u(1+r) + x_1(1+r)], N \}, \\ g_2 &= \min \left\{ \left[u(1+r)^2 + \sum_{k=1}^2 x_k(1+r)^{3-k} - \sum_{k=1}^1 y_k(1+r)^{2-k} \right], N \right\}, \\ &\dots \\ g_t &= \min \left\{ \left[u(1+r)^t + \sum_{k=1}^t x_k(1+r)^{t+1-k} - \sum_{k=1}^{t-1} y_k(1+r)^{t-k} \right], N \right\}, \end{aligned}$$

In addition, $[u(1+r) + x_1(1+r)]$ is integer part of $u(1+r) + x_1(1+r)$,

By using Lemma 2.1, $u(1+r) + x_1(1+r)$, $r(1+r)^2 + \sum_{k=1}^2 x_k(1+r)^{3-k} - \sum_{k=1}^1 y_k(1+r)^{2-k}$, ...,

$r(1+r)^t + \sum_{k=1}^t x_k(1+r)^{t+1-k} - \sum_{k=1}^{t-1} y_k(1+r)^{t-k}$ are integer numbers.

Thus, (2.10) is written as

$$\begin{aligned} B_{x_1 x_2 \dots x_t} \stackrel{as}{=} & \bigcup_{1 \leq y_1 \leq g_1} (Y_1 = y_1) \cap \left(\bigcup_{1 \leq y_2 \leq g_2} (Y_2 = y_2) \cap \dots \cap \right. \\ & \left. \cap \left(\bigcup_{1 \leq y_3 \leq g_3} (Y_3 = y_3) \cap \dots \cap \left[Y_t \leq u(1+r)^t + \sum_{k=1}^t x_k(1+r)^{t+1-k} - \sum_{k=1}^{t-1} y_k(1+r)^{t-k} \right] \dots \right) \right) \end{aligned} \quad (2.11)$$

As Assumption 2.3, we let $Y_t = y_t$ with y_t is positive integer number then $1 \leq y_t \leq N$. Combining Assumption 2.3, (2.10) and formulas define g_1, g_2, \dots, g_t , we have $1 \leq y_j \leq g_j$ ($j = \overline{1, t}$).

Therefore, (2.11) can be rearranged as

$$B_{x_1 x_2 \dots x_t} \stackrel{as}{=} \bigcup_{1 \leq y_1 \leq g_1} \bigcup_{1 \leq y_2 \leq g_2} \dots \bigcup_{1 \leq y_t \leq g_t} \{(Y_1 = y_1) \cap (Y_2 = y_2) \cap \dots \cap (Y_t = y_t)\} \quad (2.12)$$

For the reason that $Y = \{Y_n\}_{n \geq 1}$ is a sequence of independent random variables then

$$P[(Y_1 = y_1) \cap (Y_2 = y_2) \cap \dots \cap (Y_t = y_t)] = P(Y_1 = y_1) \cdot P(Y_2 = y_2) \dots P(Y_t = y_t) = q_{y_1} q_{y_2} \dots q_{y_t}$$

In the other hand, system of events $\{(Y_1 = y_1) \cap (Y_2 = y_2) \cap \dots \cap (Y_t = y_t)\}_{1 \leq y_j \leq g_j (j=1, \dots, t)}$ in (2.12) is incompatible then

$$P(B_{x_1 x_2 \dots x_t}) = \sum_{1 \leq y_1 \leq g_1} \sum_{1 \leq y_2 \leq g_2} \dots \sum_{1 \leq y_t \leq g_t} q_{y_1} q_{y_2} \dots q_{y_t} \quad (2.13)$$

Next, we consider

$$A_{x_1 x_2 \dots x_t} \stackrel{as}{=} (X_1 = x_1) \cap (X_2 = x_2) \cap \dots \cap (X_t = x_t) \\ (x_1, x_2, \dots, x_t \in E_X)$$

and

$$B_{x_1 x_2 \dots x_t} \stackrel{as}{=} \bigcup_{1 \leq y_1 \leq g_1} \bigcup_{1 \leq y_2 \leq g_2} \dots \bigcup_{1 \leq y_t \leq g_t} \{(Y_1 = y_1) \cap (Y_2 = y_2) \cap \dots \cap (Y_t = y_t)\}$$

By using, X and Y are independent, if $x_1, x_2, \dots, x_t \in E_X$ and x_1, x_2, \dots, x_n hold then $A_{x_1 x_2 \dots x_t}$ and $B_{x_1 x_2 \dots x_t}$ are independent events.

In addition, system of events $\{A_{x_1 x_2 \dots x_t} \cap B_{x_1 x_2 \dots x_t}\}_{x_1, x_2, \dots, x_t \in E_X}$ in (2.9) is incompatible.

Therefore, combining (2.8) and (2.13), we have

$$\begin{aligned} \varphi_t^{(1)}(u) &= P(A) = \sum_{x_1, x_2, \dots, x_t=1}^M P\{A_{x_1 x_2 \dots x_t} \cap B_{x_1 x_2 \dots x_t}\} = \sum_{x_1, x_2, \dots, x_t=1}^M P(A_{x_1 x_2 \dots x_t}) P(B_{x_1 x_2 \dots x_t}) \\ &= \sum_{x_1, x_2, \dots, x_t=1}^M p_{x_1} p_{x_2} \dots p_{x_t} \left(\sum_{1 \leq y_1 \leq g_1} \sum_{1 \leq y_2 \leq g_2} \dots \sum_{1 \leq y_t \leq g_t} q_{y_1} q_{y_2} \dots q_{y_t} \right) \end{aligned} \quad (2.14)$$

This completes the proof.

Corollary 2.1. Let model (1.1) satisfy Assumption 2.1 to Assumption 2.4, then finite time ruin probability of model (1.1) is defined by

$$\psi_t^{(1)}(u) = 1 - \varphi_t^{(1)}(u) = 1 - \sum_{x_1, x_2, \dots, x_t=1}^M p_{x_1} p_{x_2} \dots p_{x_t} \left(\sum_{1 \leq y_1 \leq g_1} \sum_{1 \leq y_2 \leq g_2} \dots \sum_{1 \leq y_t \leq g_t} q_{y_1} q_{y_2} \dots q_{y_t} \right) \quad (2.15)$$

Remark 2.1. Formula (2.6) (or (2.15)) gives a method to compute exactly finite time non-ruin (ruin) probability of model (1.1) which $X = \{X_n\}_{n \geq 1}$ and $Y = \{Y_n\}_{n \geq 1}$ are sequences of independent and identically distributed random variables, and they take values in a finite set of positive integer numbers.

3. Computing Ruin Probability of Generalized Risk Processes under Constant Interest Force with Homogeneous Markov Chains

Let model (1.1). We assume that:

Assumption 3.1. u, t are positive integer numbers.

Assumption 3.2. $X = \{X_n\}_{n \geq 1}$ is a homogeneous Markov chain, X_n take values in a finite set of positive integer

numbers $E_X = \{1, 2, \dots, M\}$ with $p_{ij} = P(X_{m+1} = j | X_m = i), (m \in N^* = N \setminus \{0\}, i \in E_X, j \in E_X)$ where $0 \leq p_{ij} \leq 1, \sum_{j=1}^M p_{ij} = 1$. In addition,

$$P(X_1 = i) = p_i (i \in E_X), 0 \leq p_i \leq 1, \sum_{i=1}^M p_i = 1.$$

Assumption 3.3. $Y = \{Y_n\}_{n \geq 1}$ is a homogeneous Markov chain, Y_n take values in a finite set of positive integer numbers $E_Y = \{1, 2, \dots, N\}$ with $q_{rs} = P(Y_{m+1} = s | Y_m = r), (m \in N^* = N \setminus \{0\}, r \in E_Y, s \in E_Y)$ where $0 \leq q_{rs} \leq 1, \sum_{s=1}^N q_{rs} = 1$. In addition,

$$P(Y_1 = i) = q_i (i \in E_Y), 0 \leq q_i \leq 1, \sum_{i=1}^N q_i = 1.$$

Assumption 3.4. X and Y are assumed to be independent.

Supposing that the ruin time is defined by $T_u = \inf \{j : U_j < 0\}$ where $\inf \phi = \infty$.

We define the finite time ruin (non-ruin) probability of model (1.1) using Assumption 3.1 to Assumption 3.4, respectively, by

$$\psi_t^{(2)}(u) = P(T_u \leq t) = P\left(\bigcup_{j=1}^t (U_j < 0)\right), \quad (3.1)$$

$$\phi_t^{(2)}(u) = 1 - \psi_t^{(2)}(u) = P(T_u \geq t+1) = P\left(\bigcap_{j=1}^t (U_j \geq 0)\right). \quad (3.2)$$

Similar to Theorem 2.1, we have

Theorem 3.1. Let model (1.1) satisfy Assumption 3.1 to Assumption 3.4, then finite time non-ruin probability of model (1.1) is defined by

$$\phi_t^{(2)}(u) = \sum_{x_1, x_2, \dots, x_t=1}^M p_{x_1} p_{x_1 x_2} \dots p_{x_{t-1} x_t} \left(\sum_{1 \leq y_1 \leq g_1} \sum_{1 \leq y_2 \leq g_2} \dots \sum_{1 \leq y_t \leq g_t} q_{y_1} q_{y_1 y_2} \dots q_{y_{t-1} y_t} \right), \quad (3.3)$$

where, g_1, g_2, \dots, g_t is defined in the same way with Theorem 2.1.

Proof.

We proof similarly as Theorem 2.1, where, (2.8) replaced by

$$\begin{aligned} P(A_{x_1 x_2 \dots x_t}) &= P[(X_1 = x_1) \cap (X_2 = x_2) \cap \dots \cap (X_t = x_t)] \\ &= P(X_1 = x_1) \cdot P(X_2 = x_2 | X_1 = x_1) \dots P(X_t = x_t | X_{t-1} = x_{t-1}) \\ &= p_{x_1} p_{x_1 x_2} \dots p_{x_{t-1} x_t} \end{aligned}$$

In the other hand, we have

$$\begin{aligned} &P[(Y_1 = y_1) \cap (Y_2 = y_2) \cap \dots \cap (Y_t = y_t)] \\ &= P(Y_1 = y_1) \cdot P(Y_2 = y_2 | Y_1 = y_1) \dots P(Y_t = y_t | Y_{t-1} = y_{t-1}) \\ &= q_{y_1} q_{y_1 y_2} \dots q_{y_{t-1} y_t} \end{aligned}$$

In addition, (2.13) substituted by

$$P(B_{x_1 x_2 \dots x_t}) = \sum_{1 \leq y_1 \leq g_1} \sum_{1 \leq y_2 \leq g_2} \dots \sum_{1 \leq y_t \leq g_t} q_{y_1} q_{y_1 y_2} \dots q_{y_{t-1} y_t}.$$

By using the same method to prove Theorem 2.1, we have formula (3.3).

This completes the proof.

Corollary 3.1. Let model (1.1) satisfy Assumption 3.1 to Assumption 3.4, then finite time ruin probability of model (1.1) is defined by

$$\psi_t^{(2)}(u) = 1 - \varphi_t^{(2)}(u) = 1 - \sum_{x_1, x_2, \dots, x_t=1}^M p_{x_1} p_{x_1 x_2} \dots p_{x_{t-1} x_t} \left(\sum_{1 \leq y_1 \leq g_1} \sum_{1 \leq y_2 \leq g_2} \dots \sum_{1 \leq y_t \leq g_t} q_{y_1} q_{y_1 y_2} \dots q_{y_{t-1} y_t} \right) \quad (3.4)$$

Remark 3.1. Formula (3.3) (or (3.4)) gives a method to compute exactly finite time non-ruin (ruin) probability of model (1.1) which $X = \{X_n\}_{n \geq 1}$ and $Y = \{Y_n\}_{n \geq 1}$ are homogeneous Markov chains and they take values in a finite set of positive integer numbers.

4. Numerical Illustration

4.1. Numerical Illustration for $\psi_t^{(1)}(u)$

Let $X = \{X_n\}_{n \geq 1}$ be a sequence of independent and identically distributed random variables, X_n take values in a finite set of positive integer numbers $E_X = \{1, 2, 3, 4, 5\}$ with X_1 having a distribution:

X_1	1	2	3	4	5
P	0,687918	0,107263	0,027260	0,044032	0,133522

Let $Y = \{Y_n\}_{n \geq 1}$ be a sequence of independent and identically distributed random variables, Y_n take values in a finite set of positive integer numbers $E_Y = \{1, 2, 3, 4, 5\}$ with Y_1 having a distribution:

Y_1	1	2	3	4	5
P	0,693655	0,234842	0,034024	0,022141	0,015337

By using the C program, the $\psi_t^{(1)}(u)$ is calculated with the assumptions above of random variables X_1, Y_1 .

Table 4.1 shows $\psi_t^{(1)}(u)$ for a range of value of u .

Table 4.1. Ruin probabilities of model (1.1) with Assumption 2.1- 2.4 and $r = 0,15$

	t = 3	t = 5	t = 7
u = 1,5	0,451331	0,731346	0,872909
u = 2,5	0,421682	0,720987	0,868336
u = 3,5	0,369372	0,713196	0,862931
u = 4,5	0,350304	0,703524	0,859522
u = 5,5	0,345694	0,691736	0,856952
u = 6,5	0,340705	0,687219	0,853781
u = 7,5	0,332709	0,684591	0,850476

4.2. Numerical Illustration for $\psi_t^{(2)}(u)$

Let $X = \{X_n\}_{n \geq 1}$ be a homogeneous Markov chain, X_n take values in a finite set of positive integer numbers

$E_X = \{1, 2, 3, 4, 5\}$, with X_1 having a distribution

X_1	1	2	3	4	5
P	0,412732	0,143721	0,201232	0,112731	0,129584

In addition, matrix $P = [p_{ij}]_{5 \times 5}$ is given by

$$P = \begin{bmatrix} 0,755119 & 0,169668 & 0,046277 & 0,019325 & 0,009610 \\ 0,469955 & 0,225771 & 0,074864 & 0,205732 & 0,023678 \\ 0,585528 & 0,072188 & 0,072098 & 0,241161 & 0,029025 \\ 0,376690 & 0,076737 & 0,230476 & 0,003048 & 0,313049 \\ 0,003357 & 0,621674 & 0,312923 & 0,053181 & 0,008866 \end{bmatrix}$$

Let $Y = \{Y_n\}_{n \geq 1}$ be a homogeneous Markov chain, Y_n take values in a finite set of positive integer numbers

$E_Y = \{1, 2, 3, 4, 5\}$, with Y_1 having a distribution

Y_1	1	2	3	4	5
P	0,713095	0,060022	0,118444	0,075802	0,032637

In addition, matrix $Q = [q_{ij}]_{5 \times 5}$ is given by

$$Q = \begin{bmatrix} 0,764641 & 0,105781 & 0,030568 & 0,039239 & 0,059771 \\ 0,728355 & 0,183338 & 0,031809 & 0,009616 & 0,046882 \\ 0,319773 & 0,068527 & 0,406201 & 0,199290 & 0,006209 \\ 0,422742 & 0,220847 & 0,270579 & 0,062131 & 0,023701 \\ 0,458144 & 0,073241 & 0,313488 & 0,040222 & 0,114905 \end{bmatrix}$$

By using the C program, the $\psi_t^{(2)}(u)$ is calculated with the assumptions above of random variables X_1, Y_1 and matrix P, Q .

Table 4.2 shows $\psi_t^{(2)}(u)$ for a range of value of u .

Table 4.2. Ruin probabilities of model (1.1) with Assumption 3.1-3.4 and $r = 0,15$

	t = 3	t = 5	t = 7
u = 1,5	0,438056	0,670702	0,806759
u = 2,5	0,414305	0,660951	0,802202
u = 3,5	0,387413	0,651153	0,798389
u = 4,5	0,372796	0,642734	0,795026
u = 5,5	0,368182	0,636163	0,792125
u = 6,5	0,363660	0,633552	0,789425
u = 7,5	0,357099	0,631753	0,786936

5. Conclusions

By using technique of classical probability with u, t , claims, premiums all are positive integer numbers and r is a positive number, this paper constructed an exact formula for ruin (non-ruin) probability of model (1.1), where sequences of claims and premiums are independent and identically distributed random variables or homogeneous Markov chains. Our main results in this paper not only prove Theorem 2.1 and Theorem 3.1 but also give numerical examples to illustrate for Theorem 2.1 and Theorem 3.1. These results proof for the suitability of theoretical results and practical examples. It also mean that:

When initial u is increasing then $\psi_t^{(1)}(u), \varphi_t^{(2)}(u)$ are decreasing,

With u being unchanged, when t is increasing then $\psi_t^{(1)}(u), \varphi_t^{(2)}(u)$ are increasing.

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