

# Nonlife Actuarial Models

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## Chapter 5

### Ruin Theory

# Learning Objectives

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1. Surplus function, premium rate and loss process
2. Probability of ultimate ruin
3. Probability of ruin before a finite time
4. Adjustment coefficient and Lundberg's inequality
5. Poisson process and continuous-time ruin theory

## 5.1 Discrete-Time Surplus and Ruin

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- An insurance company establishes its business with a start-up capital of  $u$  at time 0, called the **initial surplus**.
- It receives premiums of one unit per period at the end of each period. Loss claim of amount  $X_i$  is paid out at the end of period  $i$  for  $i = 1, 2, \dots$ .
- $X_i$  are independently and identically distributed as the loss random variable  $X$ .
- The **surplus** at time  $n$  with initial capital  $u$ , denoted by  $U(n; u)$ , is given by

$$U(n; u) = u + n - \sum_{i=1}^n X_i, \quad \text{for } n = 1, 2, \dots. \quad (5.1)$$

- The *numeraire* of the above equation is the amount of premium per period, or the premium rate. All other variables are measured as multiples of the premium rate.
- Thus, the initial surplus  $u$  may take values of  $0, 1, \dots$ , times the premium rate. Likewise,  $X_i$  may take values of  $j$  times the premium rate with pf  $f_X(j)$  for  $j = 0, 1, \dots$ .
- We denote the mean of  $X$  by  $\mu_X$  and its variance by  $\sigma_X^2$ .
- We assume  $X$  is of finite support, although in notation we allow  $j$  to run to infinity.

- If we denote the premium loading by  $\theta$ , then we have

$$1 = (1 + \theta)\mu_X, \quad (5.2)$$

which implies

$$\mu_X = \frac{1}{1 + \theta}. \quad (5.3)$$

We shall assume positive loading so that  $\mu_X < 1$ .

- The business is said to be in **ruin** if the surplus function  $U(n; u)$  falls to or below zero sometime after the business started, i.e., at a point  $n \geq 1$ .

**Definition 5.1:** Ruin occurs at time  $n$  if  $U(n; u) \leq 0$  for the first time at  $n$ , for  $n \geq 1$ .

**Definition 5.2:** The time-of-ruin random variable  $T(u)$  is defined as

$$T(u) = \min \{n \geq 1 : U(n; u) \leq 0\}. \quad (5.4)$$

**Definition 5.3:** Given an initial surplus  $u$ , the probability of ultimate ruin, denoted by  $\psi(u)$ , is

$$\psi(u) = \Pr(T(u) < \infty). \quad (5.5)$$

**Definition 5.4:** Given an initial surplus  $u$ , the probability of ruin by time  $t$ , denoted by  $\psi(t; u)$ , is

$$\psi(t; u) = \Pr(T(u) \leq t), \quad \text{for } t = 1, 2, \dots. \quad (5.6)$$

## 5.2 Discrete-Time Ruin Theory

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### 5.2.1 Ultimate Ruin in Discrete Time

- We now derive recursive formulas for  $\psi(u)$ .
- First, for  $u = 0$ , we have

$$\psi(0) = f_X(0)\psi(1) + S_X(0). \quad (5.7)$$

- Similarly, for  $u = 1$ , we have

$$\psi(1) = f_X(0)\psi(2) + f_X(1)\psi(1) + S_X(1). \quad (5.8)$$

- The above equations can be generalized to larger values of  $u$  as

follows

$$\psi(u) = f_X(0)\psi(u+1) + \sum_{j=1}^u f_X(j)\psi(u+1-j) + S_X(u), \quad \text{for } u \geq 1. \quad (5.9)$$

- Re-arranging equation (5.9), we obtain the following recursive formula for the probability of ultimate ruin

$$\psi(u+1) = \frac{1}{f_X(0)} \left[ \psi(u) - \sum_{j=1}^u f_X(j)\psi(u+1-j) - S_X(u) \right], \quad \text{for } u \geq 1. \quad (5.10)$$

- To apply the above equation we need the starting value  $\psi(0)$ , which is given by the following theorem.

**Theorem 5.1:** For the discrete-time surplus model,  $\psi(0) = \mu_X$ .



**Proof:** See NAM.

**Example 5.1:** The claim variable  $X$  has the following distribution:  $f_X(0) = 0.5$ ,  $f_X(1) = f_X(2) = 0.2$  and  $f_X(3) = 0.1$ . Calculate the probability of ultimate ruin  $\psi(u)$  for  $u \geq 0$ .

**Solution:** The survival function of  $X$  is  $S_X(0) = 0.2 + 0.2 + 0.1 = 0.5$ ,  $S_X(1) = 0.2 + 0.1 = 0.3$ ,  $S_X(2) = 0.1$  and  $S_X(u) = 0$  for  $u \geq 3$ . The mean of  $X$  is

$$\mu_X = (0)(0.5) + (1)(0.2) + (2)(0.2) + (3)(0.1) = 0.9,$$

which can also be calculated as

$$\mu_X = \sum_{u=0}^{\infty} S_X(u) = 0.5 + 0.3 + 0.1 = 0.9.$$

Thus, from Theorem 5.1  $\psi(0) = 0.9$ , and from equation (5.7),  $\psi(1)$  is given

by

$$\psi(1) = \frac{\psi(0) - S_X(0)}{f_X(0)} = \frac{0.9 - 0.5}{0.5} = 0.8.$$

From equation (5.8), we have

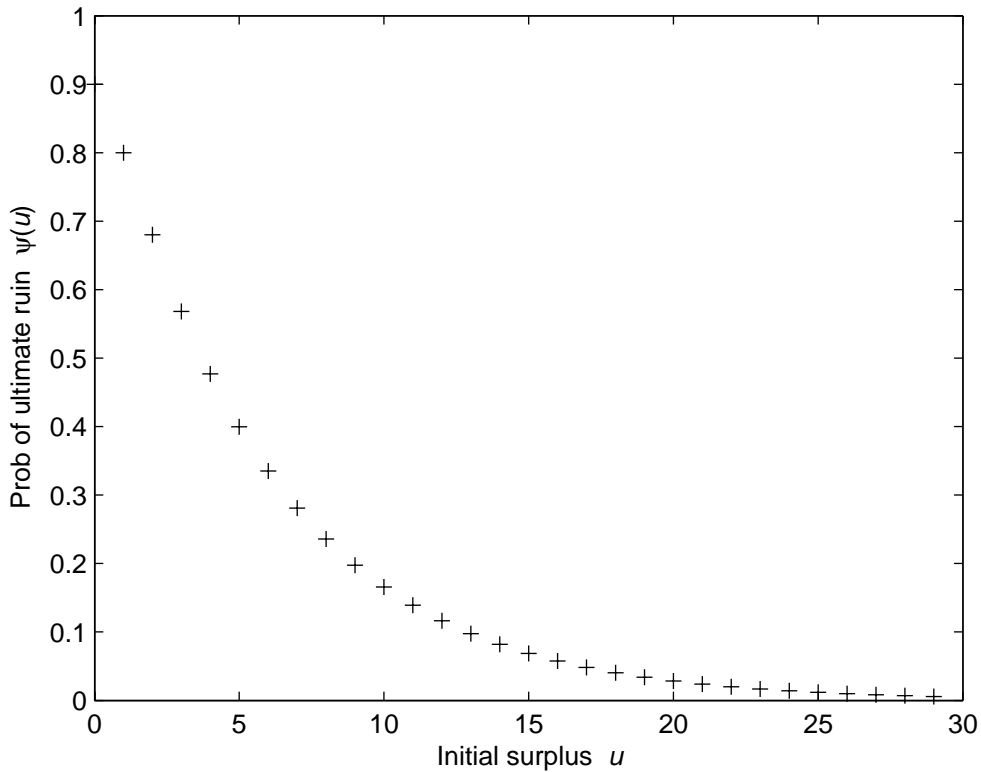
$$\psi(2) = \frac{\psi(1) - f_X(1)\psi(1) - S_X(1)}{f_X(0)} = \frac{0.8 - (0.2)(0.8) - 0.3}{0.5} = 0.68,$$

and applying equation (5.10) for  $u = 3$ , we have

$$\psi(3) = \frac{\psi(2) - f_X(1)\psi(2) - f_X(2)\psi(1) - S_X(2)}{f_X(0)} = 0.568.$$

As  $S_X(u) = 0$  for  $u \geq 3$ , using equation (5.10) we have, for  $u \geq 4$ ,

$$\psi(u) = \frac{\psi(u) - f_X(1)\psi(u) - f_X(2)\psi(u-1) - f_X(3)\psi(u-2)}{f_X(0)}.$$



### 5.2.2 Finite-Time Ruin in Discrete Time

- We now consider the probability of ruin at or before a finite time point  $t$  given an initial surplus  $u$ .
- First we consider  $t = 1$  given initial surplus  $u$ .
- As defined in equation (5.6),  $\psi(t; u) = \Pr(T(u) \leq t)$ . If  $u = 0$ , the ruin event occurs at time  $t = 1$  when  $X_1 \geq 1$ . Thus,

$$\psi(1; 0) = 1 - f_X(0) = S_X(0). \quad (5.20)$$

- Likewise, for  $u > 0$ , we have

$$\psi(1; u) = \Pr(X_1 > u) = S_X(u). \quad (5.21)$$

- We now consider  $\psi(t; u)$  for  $t \geq 2$  and  $u \geq 0$ .

- The event of ruin occurring at or before time  $t \geq 2$  may be due to (a) ruin at time 1, or (b) loss of  $j$  at time 1 for  $j = 0, 1, \dots, u$ , followed by ruin occurring within the next  $t - 1$  periods.
- When there is a loss of  $j$  at time 1, the surplus becomes  $u + 1 - j$ , so that the probability of ruin within the next  $t - 1$  periods is  $\psi(t - 1; u + 1 - j)$ .
- Thus, we conclude that

$$\psi(t; u) = \psi(1; u) + \sum_{j=0}^u f_X(j) \psi(t - 1; u + 1 - j). \quad (5.22)$$

Hence,  $\psi(t; u)$  can be computed as follows.

1. Construct a table with time  $t$  running down the rows for  $t = 1, 2, \dots$ , and  $u$  running across the columns for  $u = 0, 1, \dots$ .

2. Initialize the first row of the table for  $t = 1$  with  $\psi(1; u) = S_X(u)$ . Note that if  $M$  is the maximum loss in each period, then  $\psi(1; u) = 0$  for  $u \geq M$ .
3. Increase the value of  $t$  by 1 and calculate  $\psi(t; u)$  for  $u = 0, 1, \dots$ , using equation (5.22). Note that the computation requires the corresponding entry in the first row of the table, i.e.,  $\psi(1; u)$ , as well as some entries in the  $(t - 1)$ th row. In particular, the  $u + 1$  entries  $\psi(t - 1; 1), \dots, \psi(t - 1; u + 1)$  in the  $(t - 1)$ th row are required.
4. Re-do Step 3 until the desired time point.

**Example 5.3:** As in Example 5.1, the claim variable  $X$  has the following distribution:  $f_X(0) = 0.5$ ,  $f_X(1) = f_X(2) = 0.2$  and  $f_X(3) = 0.1$ . Calculate the probability of ruin at or before a finite time  $t$  given initial surplus  $u$ ,  $\psi(t; u)$ , for  $u \geq 0$ .

**Solution:** The results are summarized in Table 5.1 for  $t = 1, 2$  and  $3$ , and  $u = 0, 1, \dots, 6$ .

**Table 5.1:** Results of Example 5.3

Time $t$	Initial surplus $u$						
	0	1	2	3	4	5	6
1	0.500	0.300	0.100	0.000	0.000	0.000	0.000
2	0.650	0.410	0.180	0.050	0.010	0.000	0.000
3	0.705	0.472	0.243	0.092	0.030	0.007	0.001

The first row of the table is  $S_X(u)$ . Note that  $\psi(1; u) = 0$  for  $u \geq 3$ , as the maximum loss in each period is 3. For the second row, the details of the computation is as follows. First,  $\psi(2; 0)$  is computed as

$$\psi(2; 0) = \psi(1; 0) + f_X(0)\psi(1; 1) = 0.5 + (0.5)(0.3) = 0.65.$$

Similarly,

$$\psi(2; 1) = \psi(1; 1) + f_X(0)\psi(1; 2) + f_X(1)\psi(1; 1) = 0.3 + (0.5)(0.1) + (0.2)(0.3) = 0.41,$$

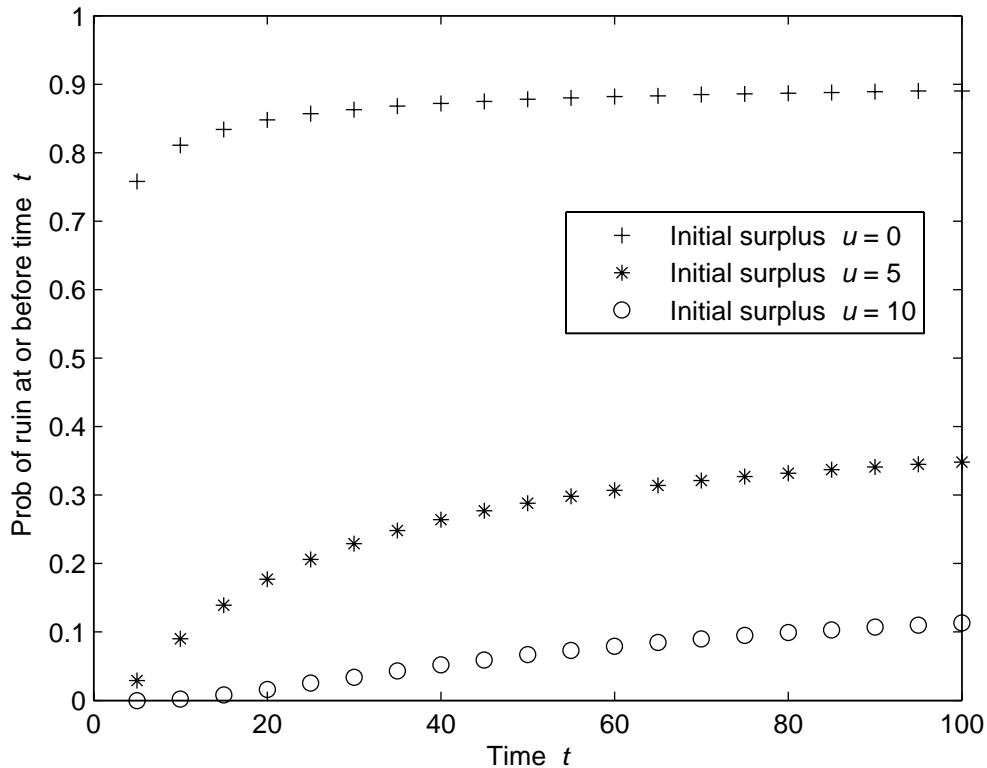
and

$$\psi(2; 2) = \psi(1; 2) + f_X(0)\psi(1; 3) + f_X(1)\psi(1; 2) + f_X(2)\psi(1; 1) = 0.18.$$

We use  $\psi(3; 3)$  to illustrate the computation of the third row as follows

$$\begin{aligned}\psi(3; 3) &= \psi(1; 3) + f_X(0)\psi(2; 4) + f_X(1)\psi(2; 3) + f_X(2)\psi(2; 2) + f_X(3)\psi(2; 1) \\ &= 0 + (0.5)(0.01) + (0.2)(0.05) + (0.2)(0.18) + (0.1)(0.41) \\ &= 0.092.\end{aligned}$$





### 5.2.3 Lundberg's inequality in Discrete Time

**Definition 5.5:** Suppose  $X$  is the loss random variable. The adjustment coefficient, denoted by  $r^*$ , is the value of  $r$  that satisfies the following equation

$$\mathbb{E} [\exp \{r(X - 1)\}] = 1. \quad (5.23)$$

**Example 5.4:** Assume the loss random variable  $X$  follows the distribution given in Examples 5.1 and 5.3. Calculate the adjustment coefficient  $r^*$ .

**Solution:** Equation (5.23) is set up as follows

$$0.5e^{-r} + 0.2 + 0.2e^r + 0.1e^{2r} = 1,$$

which is equivalent to

$$0.1w^3 + 0.2w^2 - 0.8w + 0.5 = 0,$$

for  $w = e^r$ . We solve the above equation numerically to obtain  $w = 1.1901$ , so that  $r^* = \log(1.1901) = 0.1740$ .

**Theorem 5.2 (Lundberg's Theorem):** For the discrete-time surplus function, the probability of ultimate ruin satisfies the following inequality

$$\psi(u) \leq \exp(-r^*u), \tag{5.28}$$

where  $r^*$  is the adjustment coefficient.

**Proof:** By induction, see NAM.

**Example 5.5:** Assume the loss random variable  $X$  follows the distribution given in Examples 5.1 and 5.4. Calculate the Lundberg upper bound for the probability of ultimate ruin for  $u = 0, 1, 2$  and 3.

**Solution:** From Example 5.4, the adjustment coefficient is  $r^* = 0.1740$ . The Lundberg upper bound for  $u = 0$  is 1, and for  $u = 1, 2$  and 3, we have  $e^{-0.174} = 0.8403$ ,  $e^{-(2)(0.174)} = 0.7061$  and  $e^{-(3)(0.174)} = 0.5933$ , respectively. These figures may be compared against the exact values computed in Example 5.1, namely, 0.8, 0.68 and 0.568, respectively.

## 5.3 Continuous-Time Surplus Function

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- In a continuous-time model the insurance company receives premiums continuously, while claim losses may occur at any time.
- We assume that the initial surplus of the insurance company is  $u$  and the amount of premium received per unit time is  $c$ .
- We denote the number of claims (described as the number of occurrences of events) in the interval  $(0, t]$  by  $N(t)$ , with claim amounts  $X_1, \dots, X_{N(t)}$ , which are assumed to be independently and identically distributed as  $X$ .
- We denote the aggregate losses up to (and including) time  $t$  by  $S(t)$ ,

which is given by

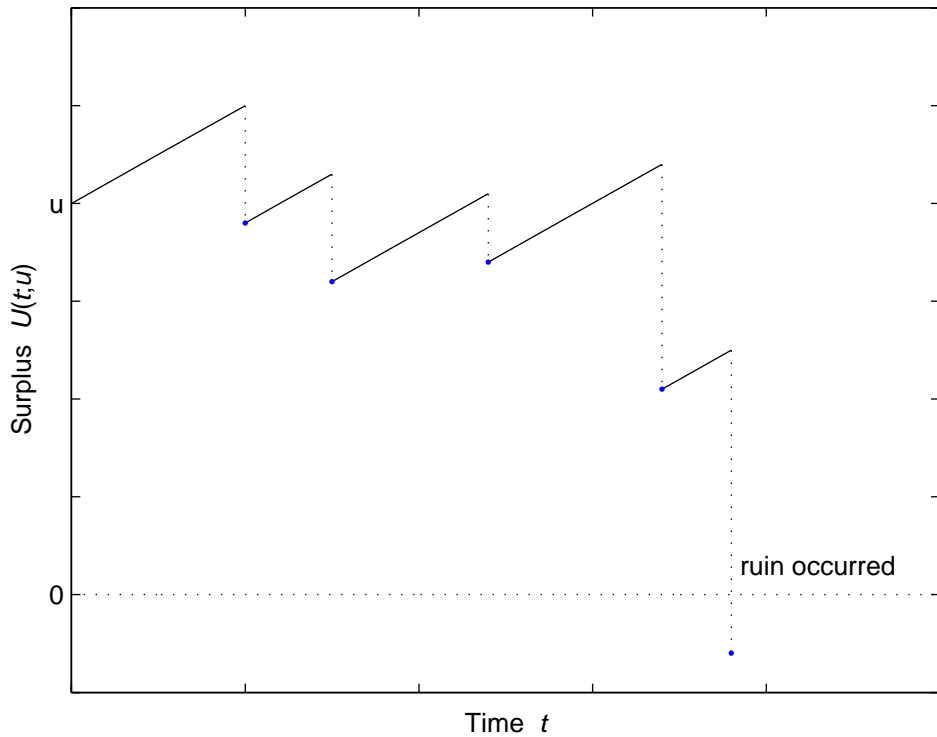
$$S(t) = \sum_{i=1}^{N(t)} X_i, \quad (5.39)$$

with the convention that if  $N(t) = 0$ ,  $S(t) = 0$ .

- Thus, the surplus at time  $t$ , denoted by  $U(t; u)$ , is defined as

$$U(t; u) = u + ct - S(t). \quad (5.40)$$

- Figure 5.4 illustrates an example of a realization of the surplus function  $U(t; u)$ .
- To analyze the behavior of  $U(t; u)$  we make some assumptions about the claim process  $S(t)$ .
- In particular, we assume that the number of occurrences of (claim) events up to (and including) time  $t$ ,  $N(t)$ , follows a **Poisson process**.



**Definition 5.6:**  $N(t)$  is a Poisson process with parameter  $\lambda$ , which is the rate of occurrences of events per unit time, if (a) in any interval  $(t_1, t_2]$ , the number of occurrences of events, i.e.,  $N(t_2) - N(t_1)$ , has a Poisson distribution with mean  $\lambda(t_2 - t_1)$ , and (b) over any non-overlapping intervals, the numbers of occurrences of events are independently distributed.

- For a *fixed*  $t$ ,  $N(t)$  is distributed as a Poisson variable with parameter  $\lambda t$ , i.e.,  $N(t) \sim \mathcal{PN}(\lambda t)$ , and  $S(t)$  follows a compound Poisson distribution.
- As a function of time  $t$ ,  $S(t)$  is a **compound Poisson process** and the corresponding surplus process  $U(t; u)$  is a **compound Poisson surplus process**. We assume that the claim random variable  $X$  has a mgf  $M_X(r)$  for  $r \in [0, \gamma)$ .



## 5.4 Continuous-Time Ruin Theory

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### 5.4.1 Lundberg's Inequality in Continuous Time

- We first define the adjustment coefficient in continuous time. Analogous to the discrete-time case, in which the adjustment coefficient is the solution of

$$1 + (1 + \theta) r \mu_X = M_X(r). \quad (5.47)$$

**Theorem 5.3:** If the surplus function follows a compound Poisson process defined in equation (5.40), the probability of ultimate ruin given initial surplus  $u$ ,  $\psi(u)$ , satisfies the inequality

$$\psi(u) \leq \exp(-r^* u), \quad (5.48)$$

where  $r^*$  is the adjustment coefficient satisfying equation (5.47).

**Example 5.6:** Let  $U(t; u)$  be a compound Poisson surplus function with  $X \sim \mathcal{G}(3, 0.5)$ . Compute the adjustment coefficient and its approximate value using equation (5.52), for  $\theta = 0.1$  and  $0.2$ . Calculate the upper bounds for the probability of ultimate ruin for  $u = 5$  and  $u = 10$ .

**Solution:** The mgf of  $X$  is, from equation (2.32),

$$M_X(r) = \frac{1}{(1 - \beta r)^\alpha} = \frac{1}{(1 - 0.5r)^3},$$

and its mean and variance are, respectively,  $\mu_X = \alpha\beta = 1.5$  and  $\sigma_X^2 = \alpha\beta^2 = 0.75$ . From equation (5.47), the adjustment coefficient is the solution of  $r$  in the equation

$$\frac{1}{(1 - 0.5r)^3} = 1 + (1 + \theta)(1.5)r,$$

from which we solve numerically to obtain  $r^* = 0.0924$  when  $\theta = 0.1$ . The upper bounds for the probability of ultimate ruin are

$$\exp(-r^*u) = \begin{cases} 0.6300, & \text{for } u = 5, \\ 0.3969, & \text{for } u = 10. \end{cases}$$

When the loading is increased to 0.2,  $r^* = 0.1718$ , so that the upper bounds for the probability of ruin are

$$\exp(-r^*u) = \begin{cases} 0.4236, & \text{for } u = 5, \\ 0.1794, & \text{for } u = 10. \end{cases}$$

To compute the approximate values of  $r^*$ , we use equation (5.52) to obtain, for  $\theta = 0.1$ ,

$$r^* \simeq \frac{(2)(0.1)(1.5)}{0.75 + (1.1)^2(1.5)^2} = 0.0864,$$

and, for  $\theta = 0.2$ ,

$$r^* \simeq \frac{(2)(0.2)(1.5)}{0.75 + (1.2)^2(1.5)^2} = 0.1504.$$

Based on these approximate values, the upper bounds for the probability of ultimate ruin are, for  $\theta = 0.1$ ,

$$\exp(-r^*u) = \begin{cases} 0.6492, & \text{for } u = 5, \\ 0.4215, & \text{for } u = 10. \end{cases}$$

and, for  $\theta = 0.2$ ,

$$\exp(-r^*u) = \begin{cases} 0.4714, & \text{for } u = 5, \\ 0.2222, & \text{for } u = 10. \end{cases}$$

Thus, we can see that the adjustment coefficient increases with the premium loading  $\theta$ . Also, the upper bound for the probability of ultimate ruin decreases with  $\theta$  and  $u$ . We also observe that the approximation of  $r^*$  works reasonably well.