



**Far Western Review**  
**A Multidisciplinary, Peer Reviewed Journal**  
ISSN: 3021-9019  
DOI: <https://doi.org/10.3126/fwr.v2i1.70533>  
Published by Far Western University  
Mahendranagar, Nepal

## Computational Methods of Ruin Probability: Actuarial Comparison of De-Vylder and Tijim's Models

**Gbenga Michael Ogungbenle<sup>1\*</sup>, Solomon Adelaja<sup>2</sup> and Alfred Timzing Chakfa<sup>3</sup>**

<sup>1,3</sup>Department of Actuarial Science, Faculty of Management Sciences, University of Jos,  
Nigeria

<sup>2</sup>Department of Mathematics, Faculty of Natural Sciences University of Jos, Nigeria.

\*Corresponding author email: moyosiolorun@gmail.com

### Abstract

The underwriting operation of insurance firms is to assume the risk of the insured in return of premium received. In order to shield itself against extreme losses and avoid the risk of insolvency, it becomes necessary to examine how the portfolio is expected to perform over a long time horizon. The surplus process is connected with the excess of the premium received over claims outgo of the insurer's portfolio to enable it predict the level at which the insurer could survive. When the surplus approaches a defined lower limit irrespective of the initial reserve, then the insurer is ruined. The probability of ruin apparently defines the volatility embedded in underwriting process as a useful tool of risk measurement in long range planning of premium rate. This paper is anchored on the following objectives: solve the adjustment co-efficient using the moment generating function, compute the De-Vylder's ruin, compute the Tijim's ruin approximation and then compare the two. Although it was verified that as the initial capital increases, the probability of ruin decreases under the two models, computational evidence from our results in tables 2 and 3 shows that the Tijim's approximation to ruin probability is higher than the De-Vylder's approximation at the same level of initial capital and safety loading. The implication is that the De-Vylder's approximation to ruin probabilities is an improvement over Tijim's ruin model and hence De-Vylder's approximation is recommended for the insurer's ruin assessment.

**Keywords:** Adjustment coefficient, minimum capital, ruin probability, surplus process, underwriting operations

Copyright 2024 © Author(s) This open access article is distributed under a Creative Commons



Attribution-Non Commercial 4.0 International (CC BY-NC 4.0) License

## Introduction

When investigating the degree of the risk connected with a portfolio of insurance policies, it becomes necessary to examine the expected performance of the portfolio over a long time horizons. In anticipation of the potential severity, the insurer usually sinks a reserve amount from which fund could be drawn as claim arises. In practice, the reserve is accumulated over a long time horizon either from the excess of premiums received or from the investment income. The surplus process is a useful analytical framework for examining how an insurer's capital evolves over time horizon. Consequently, ruin theory is connected with the excess of the premiums received over claims paid in modelling the surplus process. The surplus process  $Z(\xi)$  commences with an initial capital  $u$  and it is modelled based on both premium income per unit time and aggregate claim amount which are respectively defined in terms of cash in-flows and the cash out-flows up to time  $\xi$  and this accounts for the reason why the classical risk process is defined as a function of the initial surplus, premiums and claims. Ruin therefore arises whenever the insurer's surplus approaches a specified lower bound net of the initial reserve.

Centeno (1986) derived the insurer's adjustment coefficient as a function of the retention levels under the assumption that the annual claim is distributed as compound Poisson. Liang and Guo (2007) obtained the insurer's surplus process using the Brownian motion where closed form expression was derived in the diffusion approximation conditions to obtain an optimal strategy in minimizing the ruin probability by maximizing the adjustment coefficient. However, under the invariance hypothesis of Brownian motion Korzeniowski (2023) obtained a closed form expression of ruin probability under the discrete time risk model with random premiums.

Guerra and Centeno (2008) examined the adjustment coefficient of an arbitrary optimal reinsurance schemes on the basis of expected utility. The author further obtained a functional relationship between maximizing the adjustment coefficient and maximizing the expected utility of wealth. Burnecki, Teuerle, Wilkowska (2019) experimented the De-Vylder estimations of the ruin probability for a two-dimensional surplus process where claims and premiums are divided under a predetermined percentage. Burnecki, Mista and Weron (2003) developed a generalized De-Vylder approximation. The rationale behind this technique is to develop an advanced model for the ruin probability based on gamma claims and matching first four moments.

Santana & Rincon (2023) obtained a new model for the ultimate ruin probability in the Cramer–Lundberg risk process where the claims assume a finite mixture of  $k$  Erlang distributions. Employing the power of recurrence sequences, the technique suggested here resulted in computing ruin probability using an associated characteristic polynomial and its roots. The model is obtained through a finite sum of terms one for each zero of the polynomial in order to approximate the ruin probability.

Michna (2020) derived a model for the supremum distribution of a specified positive or negative Levy processes characterized by a broken linear drift. This resulted in an alternative model for ruin probabilities given that two underwriting firms divide between them both claims and premiums in some prescribed proportions

Luesamai (2021) derived a lower bound to obtain the finite time ruin probability that converges to the ultimate ruin probability with increasing time while the upper bound is iteratively estimated as initial point. Huang, Li, Liu and Yu (2021) estimated the ruin probability in insurance risk model under stochastic premium income where the ruin probability was computed through complex Fourier series expansion methodology.

According to Cheng, Gao and Wang (2016); Karageyik and Sahin (2016), Ogungbenle (2024), the surplus expressed in terms of the insurer's risk process  $(U(\xi))_{\xi \geq 0}$  is expressed as

$$U(\xi) = u + \pi\xi - Z(\xi) \quad (1)$$

$u$  is the initial capital,  $\pi$  is the premium per unit time paid by the insured and  $Z(\xi)$  is the aggregate claim advised against the insurer.

$$(2)$$

$$Z(\xi) = \sum_{i=1}^{m(\xi)} X_i$$

$X_i$  is the  $i$ th claim and  $M(\xi)$  is a poisson process. Consequently, we can obtain then obtain the conditional expectation of the total claim size as follows in equation (3) where  $\mathbf{E}$  and  $\mathbf{P}$  are expectation operator and probability function. Observe that

$$\mathbf{E}(Z(\xi)) = \mathbf{E}[Z(\xi)|m(\xi) = m] = m\mu \quad (3)$$

$$\mathbf{E}(Z(\xi)) = \sum_{m=0}^{\infty} (\mathbf{E}[Z(\xi)|m(\xi) = m]) \times \mathbf{P}(m(\xi) = m) \quad (3a)$$

Putting equation (3) in (3a), we have

$$\mathbf{E}(Z(\xi)) = \sum_{m=0}^{\infty} (m\mu \times \mathbf{P}(m(\xi) = m)) \quad (4)$$

Observe that the probability mass function of  $m(\xi)$  is Poisson and hence we have

$$\mathbf{P}(m(\xi) = m) = \frac{(\lambda\xi)^m e^{-\lambda\xi}}{m!} \quad (5)$$

Applying equation (4) we obtain the expectation (the first moment) of the total claim size as follows

$$\mathbf{E}(Z(\xi)) = \sum_{m=0}^{\infty} m\mu \times \frac{(\lambda\xi)^m e^{-\lambda\xi}}{m!} \quad (6)$$

$$\mathbf{E}(Z(\xi)) = \mu(\lambda\xi) e^{-\lambda\xi} \sum_{m=1}^{\infty} \frac{(\lambda\xi)^{m-1}}{(m-1)!} \quad (7)$$

Let  $Y = m - 1$  in equation (7), then we have

$$\mathbf{E}(Z(\xi)) = \mu(\lambda\xi) e^{-\lambda\xi} \sum_{m=1}^{\infty} \left( \frac{(\lambda\xi)^Y}{Y!} \right) \quad (8)$$

But observe that by definition

$$e^u = \sum_{m=0}^{\infty} \frac{u^m}{m!} \quad (9)$$

Consequently, equation (8) becomes

$$\mathbf{E}[Z(\xi)] = \mu e^{-\lambda\xi} \times \lambda\xi e^{\lambda\xi} \quad (10)$$

Therefore,

$$\mathbf{E}[Z(\xi)] = \mu \lambda \xi \quad (11)$$

We can now obtain the second moment as follows

$$\mathbf{E}[Z^2(\xi)] = \mathbf{E}\left[\left(\sum_{m=1}^{m(\xi)} X_k\right)^2\right] \quad (12)$$

By the law of total expectation

$$\mathbf{E}[Z^2(\xi)] = \sum_{m=0}^{\infty} (\mathbf{E}[Z^2(\xi)|m(\xi) = m]) \times \mathbf{P}(m(\xi) = m) \quad (13)$$

Observe that  $Z^2(\xi)$  is the squared aggregate of  $m$  independently and identically distributed random claims

$$Z^2(\xi) = (X_1 + X_2 + X_3 + \dots + X_m)^2 \quad (14)$$

$$Z^2(\xi) = X_1^2 + X_2^2 + X_3^2 + \dots + X_m^2 + \dots + 2X_1X_2 + 2X_1X_3 + \dots \quad (15)$$

Therefore

$$\mathbf{E}[Z^2(\xi)|m(\xi) = m] = \mathbf{E}(X_1^2) + \mathbf{E}(X_2^2) + \mathbf{E}(X_3^2) + \dots + \mathbf{E}(2X_1X_2) + \mathbf{E}(2X_1X_3) + \dots \quad (15)$$

Since the random losses are independently and identically distributed, we have

$$\mathbf{E}(2X_1X_2) = 2\mathbf{E}(X_1)\mathbf{E}(X_2) \quad (15a)$$

and  
 $\mathbf{E}(X) = \mathbf{E}(X_1) = \mathbf{E}(X_2) = \mathbf{E}(X_3) = \dots = \mathbf{E}(X_m)$

Therefore,

$$\mathbf{E}(X_1 X_2) = \mathbf{E}(X_2 X_3) = \mathbf{E}(X_1^2) = \mathbf{E}(X_2^2) = \dots = \mathbf{E}(X_m^2) = \mathbf{E}(X^2) \quad (17)$$

$$\mathbf{E}(X_1 X_2) = \mathbf{E}(X_1) \mathbf{E}(X_2) = \mu^2 \quad (18)$$

$$\mathbf{E}(X^2) = \mathbf{E}(X_1^2) = \mathbf{E}(X_2^2) = \mathbf{E}(X_3^2) = \dots = \mathbf{E}(X_m^2) \quad (19)$$

$$\mathbf{E}[Z^2(\xi)|m(\xi)=m] = m\mathbf{E}(X^2) + m(m-1)(\mathbf{E}(X))^2 \quad (20)$$

$$\mu = \mathbf{E}(X^2) \quad (21)$$

$$\mathbf{E}[Z^2(\xi)|m(\xi)=m] = m\mu + m(m-1)\mu^2 \quad (22)$$

$$\mathbf{E}[Z^2(\xi)|m(\xi)=m] = m\mu + m(m-1)\mu^2 \quad (23)$$

$$\mu = \sigma^2 + \mu^2 \quad (24)$$

$$\mathbf{E}[Z^2(\xi)|m(\xi)=m] = m(\sigma^2 + \mu^2) + m(m-1)\mu^2 \quad (25)$$

$$\mathbf{E}[Z^2(\xi)|m(\xi)=m] = m\sigma^2 + m\mu^2 + m^2\mu^2 - m\mu^2 \quad (26)$$

$$\mathbf{E}[Z^2(\xi)|m(\xi)=m] = m\sigma^2 + m^2\mu^2 \quad (27)$$

$$\mathbf{E}[Z^2(\xi)] = \left\{ \sum_{m=0}^{\infty} \mathbf{E}[Z^2(\xi)|m(\xi)=m] \right\} \times f_m(\xi) \quad (28)$$

$$\mathbf{E}[Z^2(\xi)] = \left\{ \sum_{m=0}^{\infty} \mathbf{E}[Z^2(\xi)|m(\xi)=m] \right\} \times \mathbf{P}(m(\xi)=m) \quad (29)$$

Equation (4) is a density function and can be rewritten as

$$f_m(\xi) = \frac{(\lambda\xi)^m e^{-\lambda\xi}}{m!} \quad (30)$$

Then at the extended point  $(\xi + s)$ , we have

$$f_m(\xi + s) = \sum_{r=0}^m f_r(\xi) f_{m-r}(s), \quad m = 0, 1, 2, \dots \quad (31)$$

But

$$\sum_{m=0}^{\infty} \mathbf{E}[Z^2(\xi)|m(\xi)=m] = \{m\sigma^2 + m^2\mu^2\} \quad (31a)$$

Substituting (31a) in (29), we have

$$\mathbf{E}[Z^2(\xi)] = \sum_{m=0}^{\infty} \left\{ \{m\sigma^2 + m^2\mu^2\} \times \frac{(\lambda\xi)^m e^{-\lambda\xi}}{m!} \right\} \quad (32)$$

Simplifying equation (32) further, we have

$$\mathbf{E}[Z^2(\xi)] = \sum_{m=0}^{\infty} \left( m\sigma^2 \frac{(\lambda\xi)^m e^{-\lambda\xi}}{m!} \right) + \sum_{m=0}^{\infty} \left( m^2\mu^2 \frac{(\lambda\xi)^m e^{-\lambda\xi}}{m!} \right) \quad (33)$$

$$\mathbf{E}[Z^2(\xi)] = \sum_{m=0}^{\infty} \left( \sigma^2 \frac{(\lambda\xi)^m e^{-\lambda\xi}}{(m-1)!} \right) + \sum_{m=0}^{\infty} \left( m\mu^2 \frac{(\lambda\xi)^m e^{-\lambda\xi}}{(m-1)!} \right) \quad (34)$$

Now the first two moments of the counting process  $\mathbf{E}(m(\xi))$  are given in equations (35) and (35a) to enable us obtain the first two moments of the total claim size

$$\mathbf{E}(m(\xi)) = \sum_{m=0}^{\infty} \left( \frac{(\lambda\xi)^m e^{-\lambda\xi}}{(m-1)!} \right) \quad (35)$$

$$\mathbf{E}[m^2(\xi)] = \sum_{m=0}^{\infty} \left( m\mu^2 \frac{(\lambda\xi)^m e^{-\lambda\xi}}{(m-1)!} \right) \quad (35a)$$

$$\mathbf{E}[Z^2(\xi)] = \sigma^2 \mathbf{E}(m(\xi)) + \mu^2 \mathbf{E}(m^2(\xi)) \quad (36)$$

$$\mathbf{E}[Z^2(\xi)] = \sigma^2 \mathbf{E}(m(\xi)) + \mu^2 \left\{ VAR(m(\xi)) + [\mathbf{E}(m(\xi))]^2 \right\} \quad (37)$$

Again, observe that

$$(m(\xi)) = VAR(m(\xi)) = \lambda \xi \quad (38)$$

$$\mathbf{E}[Z^2(\xi)] = \sigma^2 \lambda \xi + \mu^2 \{\lambda \xi + \lambda^2 \xi^2\} \quad (39)$$

$$\mathbf{E}[Z^2(\xi)] = \sigma^2 \lambda \xi + \lambda \xi \mu^2 + \lambda^2 \xi^2 \mu^2 = \lambda \xi (\sigma^2 + \mu^2 + \lambda \xi \mu^2) \quad (40)$$

$$\mathbf{E}[Z^2(\xi)] = \sigma^2 \lambda \xi + \lambda \xi \mu^2 + \lambda^2 \xi^2 \mu^2 = \lambda \xi (\mu_2 + \lambda \xi \mu^2) \quad (41)$$

So that the expected value of the whole of the surplus process in equation (1) becomes

$$\mathbf{E}(U(\xi)) = u + \pi \xi - \mathbf{E}(Z(\xi)) \quad (42)$$

$$\mathbf{E}(U(\xi)) = u + \pi \xi - \mu \lambda \xi \quad (43)$$

The probability of the ultimate ruin in continuous time for infinite time is expressed as  $\psi(u) = \mathbf{P}(u(\xi) < 0 \text{ for some } \xi > 0)$  and represents the probability that the insurer's surplus declines below a prescribed time in the long run and such that the claims size outgo exceed the initial surplus plus premium income. In the classical risk process,  $m(\xi)$  is a homogeneous Poisson process with intensity  $\lambda$  while  $U(\xi)$  represents the classical risk model or Cramer-Lundberg model. The Cramer's condition is obtained by applying the Esscher transform

$$\int_0^{Rx} e^{Rx} (S_X(x)) dx = \mu(1+\theta) \quad (44)$$

Observe that  $e^{Rx} (S_X(x)) \rightarrow 0$  as  $x \rightarrow \infty$  and  $S_X(x) \rightarrow 0$  faster than  $e^{-Rx}$  making the tail of  $1 - F_X(x)$  to become light. We observe a counter example for the Pareto density

$$f_X(x) = \alpha \frac{\beta^\alpha}{x^{\alpha+1}} \quad (44a)$$

The tail of Pareto distribution is

$$1 - F_X(x) = \left( \frac{\beta}{x} \right)^\alpha \quad (44b)$$

Therefore,  $e^{Rx} (1 - F_X(x)) = e^{Rx} \left( \frac{\beta}{x} \right)^\alpha \rightarrow \infty$  for  $R > 0$  and hence the Lundberg exponent does not exist and consequently the above counter example implies that ruin probability can violate the Lundberg's upper bound particularly for large initial capital. In a risk process where claim size are  $X \geq 0$  and  $\mathbf{E}(X) = \mu > 0$ , the adjustment equation has a trivial solution but the adjustment coefficient  $R$  is a positive real number satisfying the equation

$$M_X(R) - 1 = (1+\theta) \mu(R) \quad (45a)$$

The adjustment coefficient is then solved from

$$\lambda + \pi R = \lambda M_X(R) \quad (45c)$$

Suppose  $Z$  is the total claim size in an interval of length 1, then  $(\pi - Z)$  defines the profit in that interval.

We assume further that  $Z$  is a compound Poisson with intensity  $\lambda$  and the moment generating function of  $Z$  is  $M_{Z_\xi}(R) = M_{m_\xi}(\log_e M_X(R)) = e^{\lambda \xi (M_X(R)-1)}$

Differentiating equation (45b) once, we obtain

$$M_{Z_\xi}'(R) = \lambda \xi M_X'(R) e^{\lambda \xi (M_X(R)-1)} \quad (45c)$$

From (45c),

$$M_{X'}(0) = 1 \quad (45d)$$

$$M_{Z_\xi}(0) = \lambda \xi M_X'(0) = \lambda \xi \mu \quad (46)$$

The equation (46) could be computationally prohibitive to solve but it can be employed to derive an inequality.

For compound Poisson process, the upper bound for the infinite time ruin probability yields that

$$\psi(u) \leq e^{-Ru} \quad (47)$$

The Cramer-Lundberg's asymptotic infinite time ruin probability model for large value of initial capital  $u$  is given as  $\psi(u) \square Be^{\theta u}$  where

$$B = \frac{\theta\mu}{M'_x(R) - \mu(1 + \theta)} \quad (48)$$

The ruin probability decreases quickly when the adjustment coefficient increases. The Cramer-Lundberg's model requires that the adjustment coefficient exists and assumes light-tailed distributions. However if the individual claim value is exponentially distributed with the parameter  $\alpha$ , the precise infinite time ruin probability will be obtained as  $\psi(u) = \psi(0)e^{-Ru}$  where  $\psi(0) = \frac{1}{(1 + \theta)}$  and  $R = \alpha - \frac{\lambda}{\pi}$

### The Adjustment Coefficient

Suppose  $\mathbf{P}$  defines a probability measure with

$$\mathbf{E}(\mathbf{P}) = \int x d\mathbf{P}(x) \in ]-\infty, 0[ \quad (48a)$$

The moment generating function defined by  $M_p(\beta) = \int e^{\beta x} d\mathbf{P}(x)$  for finite  $M_p(\beta) < \infty$  for some  $\alpha > 0$  and  $\lim_{\beta \rightarrow \xi} M_p(\beta) \geq 1$  for  $\xi = \sup \{\beta > 0 : M_p(\beta) < \infty\}$ . If  $\exists$  a real number

$R > 0$  such that  $M_p(R) = 1$ , then  $R$  is the adjustment coefficient of  $\mathbf{P}$ . The adjustment coefficient  $R$  exists under the following conditions

$M_p(0) = 1$ ,  $M^{(1)}_p(0) < 0$  and  $M_p(\beta)$  is strictly convex as  $M^{(2)}_p(\beta) > 0$  for all  $\beta < \xi$ . The number  $R$  is the only positive number such that  $e^{Rx} d\mathbf{P}(x)$  is a probability measure.

### Lundberg Inequality

Let  $(Z_n : n \in \mathbf{N})$  define a sequence of independent and identical random losses with distribution  $P$

$$\Psi(u) = \mathbf{P}\left(\sup\left\{\sum_{i=1}^m Z_i : m \in \mathbf{N}\right\} > u\right) \quad (49)$$

for all  $u \geq 0$

Define

$$\Psi_n(u) = \begin{cases} \mathbf{P}\left(\sup\left\{\sum_{i=1}^m Z_i : m \leq n\right\} > u\right), & u \geq 0 \\ 1, & u < 0 \end{cases} \quad (50)$$

For all  $n \in \mathbf{N}$ , we show that  $\Psi_n(u) \leq e^{-Ru}$  for all  $u \in \mathbf{R}$  by inductive principles on  $n$ . Observe that

$$\Psi(u) = \lim_{n \rightarrow \infty} \Psi_n(u) \quad (51)$$

For  $u \leq 0$  and all  $n \in \mathbf{N}$ , the bound is true from definition. Let  $n = 1$  and  $u > 0$ . Then

$$\Psi_1(u) = \mathbf{P}(Z_1 > u) = \mathbf{P}(RZ_1 > Ru) \quad (52)$$

$$\Psi_1(u) \leq e^{-Ru} \mathbf{E}(e^{RZ_1}) = e^{-Ru} \quad (53)$$

By the Independence assumption of  $Z_n$ , we have

$$\Psi_{n+1}(u) = 1 - \int_{-\infty}^u 1 - \Psi_n(u-z) d\mathbf{P}(z) \quad (54)$$

$$\Psi_{n+1}(u) = 1 - P([- \infty, u]) + \int_{-\infty}^u \Psi_n(u-z) d\mathbf{P}(z) \quad (55)$$

$$\Psi_{n+1}(u) = \int_{-\infty}^{\infty} \Psi_n(u-z) d\mathbf{P}(z) \quad (56)$$

$$\Psi_{n+1}(u) \leq \int_{-\infty}^{\infty} e^{-R(u-z)} d\mathbf{P}(z) = e^{-Ru} \quad (57)$$

### Theorem

$$\text{If } \frac{\pi}{\alpha\mu} \leq \frac{(e^{RK} - 1)}{RK} \quad (57a)$$

then

$$\pi e^{-RK} \leq \pi - \frac{\pi}{2} RK + \frac{\pi}{12} R^2 K^2 - \frac{\pi}{720} R^4 K^4 < \alpha \mu \quad (58)$$

**Proof**

$$\frac{Y}{K} e^{RK} + \left(1 - \frac{Y}{K}\right) = 1 - \frac{Y}{K} + \frac{Y}{K} + YR + \frac{YR^2 K}{2!} + \frac{YR^3 K^2}{3!} + \frac{YR^4 K^3}{4!} + \dots \quad (59)$$

By Jensen's inequality,

$$E(f(Y)) \leq f(E(Y)) \quad (60)$$

$$\alpha + \pi R \leq \alpha E(e^{RY}) \quad (61)$$

$$\alpha + \pi R \leq \alpha E\left(\frac{Y}{K} e^{RK} + \left(1 - \frac{Y}{K}\right)\right) \quad (62)$$

$$\alpha + \pi R \leq \alpha E\left(\frac{Y}{K} e^{RK}\right) + \alpha E\left(1 - \frac{Y}{K}\right) \quad (63)$$

$$\alpha + \pi R \leq \alpha E\left(\frac{Y}{K}\right) e^{RK} + \alpha \left(1 - \frac{E(Y)}{K}\right) \quad (64)$$

$$\alpha + \pi R \leq \alpha \left(\frac{\mu}{K}\right) e^{RK} + \alpha \left(1 - \frac{\mu}{K}\right) \quad (65)$$

$$\alpha + \pi R \leq \frac{\alpha \mu}{K} e^{RK} + \alpha - \frac{\alpha \mu}{K} \quad (66)$$

$$\pi R \leq \alpha \frac{\mu}{K} (e^{RK} - 1) \quad (67)$$

$$\frac{(e^{RK} - 1)}{KR} \geq \frac{\pi}{\alpha \mu} \quad (68)$$

Observe that the following inequality holds

$$\frac{(e^{RK} - 1)}{RK} < e^{RK} \quad (69)$$

Implying

$$\frac{\pi}{\alpha \mu} \leq \frac{(e^{RK} - 1)}{RK} < e^{RK} \quad (70)$$

Taking reciprocals of all sides of the inequalities, we have

$$\frac{\alpha \mu}{\pi} \geq \frac{RK}{(e^{RK} - 1)} > e^{-RK} \quad (71)$$

Applying the Euler-Maclaurin series formula (71a)

$$\frac{y}{e^y - 1} = \sum_{j=0}^{\infty} \frac{B_j x^j}{j!} \quad (71a)$$

where  $B_j$  are the rational Bernoulli numbers

$$\frac{y}{e^y - 1} \approx 1 - \frac{y}{2} + \frac{y^2}{12} - \frac{y^4}{720} + \frac{y^6}{30240} - \frac{y^8}{1209600} + \frac{y^{10}}{47900160} \quad (71b)$$

we have

$$\frac{\alpha \mu}{\pi} \geq 1 - \frac{1}{2} RK + \frac{1}{12} R^2 K^2 - \frac{1}{720} R^4 K^4 > e^{-RK} \quad (72)$$

*QED*

**Ruin Probability when the Claim Size is Exponentially Distributed.**

Suppose the claim size is exponentially distributed that is  $X \sim EXP(\alpha^-)$ , then in order to obtain the probability of ruin, it is easier to consider the non-ruin integral equation as follows

$$\frac{d\phi(u)}{du} = \frac{\lambda}{\pi} \phi(u) - \frac{\lambda}{\pi} \int_0^u \phi(u-x) f_X(x) dx \quad (73)$$

$$Z = u - x \Rightarrow dZ = -dx \quad (74)$$

$$\frac{d\phi(u)}{du} = \frac{\lambda}{\pi} \phi(u) - \frac{\lambda}{\alpha\pi} \int_0^u \phi(x) e^{-\frac{(u-x)}{\alpha}} dx \quad (75)$$

$$\frac{d\phi(u)}{du} = \frac{\lambda}{\pi} \phi(u) - \frac{\lambda}{\alpha\pi} e^{-\frac{u}{\alpha}} \int_0^u \phi(x) e^{\frac{x}{\alpha}} dx \quad (76)$$

$$\frac{d^2\phi(u)}{du^2} = \frac{\lambda}{\pi} \frac{d\phi(u)}{du} + \frac{1}{\alpha} \left( \frac{\lambda}{\pi} \phi(u) - \frac{d\phi(u)}{du} \right) - \frac{\lambda}{\pi\alpha} \phi(u) \quad (77)$$

$$\frac{d^2\phi(u)}{du^2} = \frac{\lambda}{\pi} \frac{d\phi(u)}{du} - \frac{1}{\alpha} \frac{d\phi(u)}{du} + \frac{1}{\alpha} \frac{\lambda}{\pi} \phi(u) - \frac{\lambda}{\pi\alpha} \phi(u) \quad (78)$$

$$\frac{d^2\phi(u)}{du^2} = \left( \frac{\lambda}{\pi} - \frac{1}{\alpha} \right) \frac{d\phi(u)}{du} = -\frac{\theta}{\alpha(1+\theta)} \frac{d\phi(u)}{du} \quad (79)$$

$$\frac{d^2\phi(u)}{du^2} = \left( \frac{\alpha\lambda-1}{\alpha\pi} \right) \frac{d\phi(u)}{du} = \left( \frac{\frac{\alpha\lambda}{\pi}-1}{\alpha} \right) \frac{d\phi(u)}{du} \quad (80)$$

where

$$\frac{\alpha\lambda}{\pi} = \frac{1}{1+\theta} \quad (81)$$

$$\frac{d^2\phi(u)}{du^2} = \left( \frac{\frac{1}{1+\theta}-1}{\alpha} \right) \frac{d\phi(u)}{du} = \left( \frac{\frac{1-1-\theta}{1+\theta}}{\alpha} \right) \frac{d\phi(u)}{du} \quad (82)$$

$$\frac{d^2\phi(u)}{du^2} = \left( \frac{-\theta}{\alpha(1+\theta)} \right) \frac{d\phi(u)}{du} = \frac{-\theta}{\alpha(1+\theta)} \frac{d\phi(u)}{du} \quad (83)$$

$$\frac{d^2\phi(u)}{du^2} = \frac{-\theta u}{\alpha(1+\theta)} \frac{d\phi(u)}{du} \Rightarrow \frac{\phi''(u)}{\phi'(u)} = -\frac{\theta u}{\alpha(1+\theta)} = \frac{d}{du} (\log_e \phi'(u)) \quad (84)$$

$$\log_e \left[ \frac{d\phi(u)}{du} \right] = \frac{-\theta u}{\alpha(1+\theta)} + \kappa_1 \quad (85)$$

$$\frac{d\phi(u)}{du} = \kappa_2 \exp \left( \frac{-\theta u}{\alpha(1+\theta)} \right) \Rightarrow \phi(u) = \kappa_3 \exp \left( \frac{-\theta u}{\alpha(1+\theta)} \right) + \kappa_4 \quad (86)$$

By the monotone convergence theorem, we have

$$\phi(\infty) = 1 \Rightarrow \kappa_4 = 1 \quad (86a)$$

$$\phi(0) = 1 - \frac{1}{(1+\theta)} \Rightarrow \kappa_3 = -\frac{1}{(1+\theta)} \quad (87)$$

$$\phi(u) = 1 - \frac{1}{1+\theta} e^{\left( \frac{-\theta u}{\alpha(1+\theta)} \right)} \quad (88)$$

Therefore, the ruin probability is given as

$$\psi(u) = 1 - \phi(u) = \frac{1}{1+\theta} e^{\left(\frac{-\theta u}{\alpha(1+\theta)}\right)} \quad (89)$$

$$\psi(0) = \frac{1}{1+\theta} \quad (89a)$$

**Theorem**

Suppose that the ruin probability  $\psi(u)$  is differentiable, then within the interval  $0 < u < \frac{\pi}{\delta}$  where  $0 < \delta < 1$ ,

$$(\delta u - \pi)\psi^{(1)}(u) = -\lambda \int_u^\infty S_X(\xi - u)H(\xi)d\xi \quad (90a)$$

$$\psi^{(2)}(u) + \psi^{(1)}(u) = \left[ \begin{array}{l} \lambda H(u) \\ \left. \left( + \int_u^\infty \lambda h(\xi - u)S_X(\xi - u)H(\xi) - \lambda S_X(\xi - u)H(\xi)d\xi \right) \right\} \\ \left. (\delta u - \pi) \right. \end{array} \right] \quad (90b)$$

**Proof:**

Let  $f_X(x)$ ,  $S_{X^{(1)}}(x)$  and  $F_{X^{(1)}}(x)$  be the probability density and survival functions and distribution function respectively.  $\psi^{(1)}(u)$  and  $\psi^{(2)}(u)$  are the first and second order co-efficients with respect to argument  $u$ . Observe that the ruin probability satisfies the integro-differential equation defined by

$$(\delta u - \pi)\psi'(u) = \lambda\psi(u) - \lambda \int_0^{\delta-u} f_X(x)\psi(u+x)dx \quad (91)$$

But

$$\int_0^\infty f_X(x)\psi(u)dx = \psi(u) \int_0^\infty f_X(x)dx \quad (92)$$

$$\int_0^\infty f_X(x)\psi(u)dx = \psi(u).1 = \psi(u) \quad (93)$$

again

$$\int_0^\infty f_X(x)\psi(u)dx = \int_0^{\delta-u} f_X(x)\psi(u)dx + \int_{\delta-u}^\infty f_X(x)\psi(u)dx \quad (94)$$

Therefore

$$\lambda\psi(u) - \lambda \int_0^{\delta-u} f_X(x)\psi(u+x)dx = \lambda \int_0^{\delta-u} f_X(x)\psi(u)dx - \lambda \int_0^{\delta-u} f_X(x)\psi(u+x)dx + \lambda \int_{\delta-u}^\infty f_X(x)\psi(u)dx \quad (95)$$

Let the function  $H(\cdot)$  defined in equations (90a) and (90b)

$$H(u) = \begin{cases} \frac{d}{du}\psi(u) & \text{for } 0 \leq u \leq \frac{\pi}{\delta} \\ 0 & \text{for } u > \frac{\pi}{\delta} \end{cases} \quad (96)$$

Recall that the initial capital satisfies the inequality  $0 \leq u \leq \frac{\pi}{\delta}$ , consequently we have

$$\lambda\psi(u) - \lambda \int_0^{\delta} f_X(x)\psi(u+x)dx = \lambda \int_0^\infty f_X(x) \int_{u+x}^{\delta} H(\xi)d\xi dx \quad (97)$$

$$\lambda\psi(u) - \lambda \int_0^{\delta} f_X(x)\psi(u+x)dx = \lambda \int_0^\infty f_X(x) \int_{u+x}^{\delta} \psi'(\xi)d\xi dx \quad (98)$$

$$\lambda\psi(u) - \lambda \int_0^{\delta} f_X(x)\psi(u+x)dx = \lambda \int_0^\infty f_X(x) [\psi(\xi)]_{u+x}^{\delta} dx \quad (99)$$

$$\lambda\psi(u) - \lambda \int_0^{\delta} f_X(x)\psi(u+x)dx = \lambda \int_0^\infty f_X(x) [\psi(u) - \psi(u+x)]dx \quad (100)$$

By changing the order of integration we have

$$\lambda\psi(u) - \lambda \int_0^{\frac{\pi}{\delta}} f_x(x)\psi(u+x)dx = -\lambda \int_u^{\infty} \int_{\xi=u}^{\infty} f_x(x)dx H(\xi)d\xi \quad (101)$$

$$\lambda\psi(u) - \lambda \int_0^{\frac{\pi}{\delta}} f_x(x)\psi(u+x)dx = -\lambda \int_u^{\infty} S_x(\xi-u)H(\xi)d\xi \quad (102)$$

$$(\delta u - \pi)\psi^{(1)}(u) = -\lambda \int_u^{\infty} S_x(\xi-u)H(\xi)d\xi \quad (103)$$

$$(\delta u - \pi)\psi^{(2)}(u) = -\lambda \left[ -S_x(0)H(u) + \int_u^{\infty} \frac{\partial}{\partial u} S_x(\xi-u)H(\xi)d\xi \right] \quad (104)$$

$$(\delta u - \pi)\psi^{(2)}(u) = \left[ \lambda H(u) - \lambda \int_u^{\infty} \frac{\partial}{\partial u} S_x(\xi-u)H(\xi)d\xi \right] \quad (105)$$

$$(\delta u - \pi)\psi^{(2)}(u) = \lambda H(u) - \lambda \int_u^{\infty} S'_x(\xi-u)H(\xi)d\xi \quad (106)$$

By reason of the differential co-efficient under the integral operator in equations (104), (105) and (106), we differentiate the distribution function. Therefore, differentiating the distribution function  $F_x(x)$ , we have the probability density function  $f_x(x)$ ,

$$\frac{d}{dx} F_x(x) = f_x(x) \quad (107)$$

Recall that the survival and the distribution function of the claim size  $X$  add up to 1

$$S_x(x) + F_x(x) = 1 \quad (108)$$

Therefore, using equation (108), we obtain

$$\frac{d}{dx} S_x(x) + f_x(x) = 0 \quad (109)$$

$$\frac{d}{dx} S_x(x) = -f_x(x) \quad (110)$$

Substituting equation (110) in equation (106), we obtain

$$(\delta u - \pi)\psi^{(2)}(u) = \lambda H(u) - \lambda \int_u^{\infty} (-1)f_x(\xi-u)H(\xi)d\xi \quad (111)$$

$$(\delta u - \pi)\psi^{(2)}(u) = \lambda H(u) + \lambda \int_u^{\infty} f_x(\xi-u)H(\xi)d\xi \quad (112)$$

$$(\delta u - \pi)\psi^{(2)}(u) + (\delta u - \pi)\psi^{(1)}(u) = \lambda H(u) \\ + \lambda \int_u^{\infty} f_x(\xi-u)H(\xi)d\xi - \lambda \int_u^{\infty} S_x(\xi-u)H(\xi)d\xi \quad (113)$$

$$\psi^{(2)}(u) + \psi^{(1)}(u) = \frac{1}{(\delta u - \pi)} \left[ \lambda H(u) + \lambda \int_u^{\infty} f_x(\xi-u)H(\xi)d\xi - \lambda \int_u^{\infty} S_x(\xi-u)H(\xi)d\xi \right] \quad (114)$$

But the hazard function is

$$h(\xi-u) = \frac{f_x(\xi-u)}{S_x(\xi-u)} \quad (115)$$

$$\psi^{(2)}(u) + \psi^{(1)}(u) = \left[ \frac{\left\{ \begin{array}{l} \lambda H(u) \\ + \lambda \int_u^{\infty} h(\xi-u)S_x(\xi-u)H(\xi)d\xi - \lambda \int_u^{\infty} S_x(\xi-u)H(\xi)d\xi \end{array} \right\}}{(\delta u - \pi)} \right] \quad (116)$$

$$\psi^{(2)}(u) + \psi^{(1)}(u) - \left[ \frac{\left\{ \lambda H(u) + \int_u^\infty \lambda h(\xi-u) S_x(\xi-u) H(\xi) - \lambda S_x(\xi-u) H(\xi) d\xi \right\}}{(\delta u - \pi)} \right] = 0$$

*QED*

(117)

### Material and Methods

#### Numerical Estimation of the Adjustment Co-efficient

The adjustment co-efficient  $R$  in equations (45a), (47), (48) and (53) cannot be analytically determined unless it is numerically estimated to enable us estimate probability of ruin under the chosen models.

$$M_x(R) - 1 = \frac{\pi R}{\lambda} \quad (118)$$

$$\lambda + \pi R = \lambda M_x(R) \quad (119)$$

$$\lambda + \pi R = \lambda \int_0^\infty e^{Rx} f_x(x) dx \quad (120)$$

$$\lambda + \pi R = \lambda \int_0^\infty \left( 1 + Rx + \frac{R^2 x^2}{2} + \frac{R^3 x^3}{3} + \frac{R^4 x^4}{4} + \dots \right) f_x(x) dx \quad (121)$$

Ignoring the fifth and higher terms, we have

$$\lambda + \pi R \approx \lambda \int_0^\infty \left( 1 + Rx + \frac{R^2 x^2}{2} + \frac{R^3 x^3}{3} \right) f_x(x) dx \quad (122)$$

$$\lambda + \pi R \approx \lambda \left\{ \int_0^\infty f_x(x) dx + R \int_0^\infty x f_x(x) dx + \frac{R^2}{2} \int_0^\infty x^2 f_x(x) dx + \frac{R^3}{3} \int_0^\infty x^3 f_x(x) dx \right\} \quad (123)$$

$$\lambda + \pi R = \lambda \left[ 1 + R \times \mathbf{E}(X) + \frac{R^2}{2} \mathbf{E}(X^2) + \frac{R^3}{3} \mathbf{E}(X^3) \right] \quad (124)$$

$$\lambda + \pi R = \left[ \lambda + \lambda R \mathbf{E}(X) + \frac{\lambda R^2}{2} \mathbf{E}(X^2) + \frac{\lambda R^3}{3} \mathbf{E}(X^3) \right] \quad (125)$$

$$\pi R = \lambda R \mathbf{E}(X) + \frac{\lambda R^2}{2} \mathbf{E}(X^2) + \frac{\lambda R^3}{3} \mathbf{E}(X^3) \quad (126)$$

$$\pi = \lambda \mathbf{E}(X) + \frac{\lambda R}{2} \mathbf{E}(X^2) + \frac{\lambda R^2}{3} \mathbf{E}(X^3) \quad (127)$$

But

$$\pi = (1 + \theta) \lambda \mathbf{E}(X) \quad (128)$$

$$(1 + \theta) \lambda \mathbf{E}(X) = \frac{\lambda R^2}{3} \mathbf{E}(X^3) + \frac{\lambda R}{2} \mathbf{E}(X^2) + \lambda \mathbf{E}(X) \quad (129)$$

$$2\lambda \mathbf{E}(X^3) R^2 + 3\lambda \mathbf{E}(X^2) R + 6\lambda \mathbf{E}(X) - 6(1 + \theta) \lambda \mathbf{E}(X) = 0 \quad (130)$$

$$2\lambda \mathbf{E}(X^3) R^2 + 3\lambda \mathbf{E}(X^2) R + 6\lambda \mathbf{E}(X) - 6\lambda \mathbf{E}(X) - 6\lambda \theta \mathbf{E}(X) = 0 \quad (131)$$

$$2\lambda \mathbf{E}(X^3)R^2 + 3\lambda \mathbf{E}(X^2)R - 6\lambda \theta \mathbf{E}(X) = 0 \quad (132)$$

$$R = \frac{-3\lambda \mathbf{E}(X^2) \pm \sqrt{(3\lambda \mathbf{E}(X^2))^2 + 48\lambda^2 \theta \times \mathbf{E}(X) \times \mathbf{E}(X^3)}}{4\lambda \mathbf{E}(X^3)} \quad (133)$$

$$R = \frac{-3\lambda \mathbf{E}(X^2) + \sqrt{(3\lambda \mathbf{E}(X^2))^2 + 48\lambda^2 \theta \times \mathbf{E}(X) \times \mathbf{E}(X^3)}}{4\lambda \mathbf{E}(X^3)} \quad (134)$$

$$R = \frac{-3\lambda \mathbf{E}(X^2) + \sqrt{9\lambda^2 (\mathbf{E}(X^2))^2 + 48\lambda^2 \theta \times \mathbf{E}(X) \times \mathbf{E}(X^3)}}{4\lambda \mathbf{E}(X^3)} \quad (135)$$

$$R = \frac{\sqrt{9\lambda^2 \mu_2^2 + 48\lambda^2 \theta \mu \mu_3} - 3\mu_2}{4\mu_3} \quad (136)$$

**The Tijim's approximation to ruin probability**

$$\psi_T(u) = \left( \frac{1}{1+\theta} - A \right) e^{-\frac{u}{\alpha}} + Ae^{-Ru} \quad (137)$$

$$\alpha = \frac{\mathbf{E}(X^2) - A}{\frac{1}{(1+\theta)} - A} \quad (138)$$

$\theta$  = Loading and  $\mu = \mathbf{E}(X)$

Since

$$\frac{\lambda \mu}{\pi} = \frac{1}{(1+\theta)} \quad (139)$$

$$A = \frac{\mu \theta}{M'_X(R) - \mu(1+\theta)} \quad (140)$$

$A$  is from the Cramer's asymptotic approximation

### The De-Vylder's Approximation

The rationale behind this model is to replace the surplus process with one with  $\bar{\theta} = \theta$ ,  $\bar{\lambda} = \lambda$  and exponential claims with parameters  $\bar{\beta} = \beta$

$$\psi_D(u) = \frac{1}{1+\bar{\theta}} e^{-\frac{\bar{\theta}\bar{\beta}u}{1+\bar{\theta}}} \quad (141)$$

where

$$\bar{\theta} = \frac{2\mu\mu_3\theta}{3(\mu_2)^2} \quad (142)$$

$$\bar{\beta} = \frac{3(\mu_2)}{(\mu_3)} \quad (143)$$

$$\bar{\lambda} = \frac{9\lambda(\mu_2)^3}{2(\mu_3)^2} \quad (144)$$

## Data Analysis and Presentation

### The Adjustment Coefficient

**Table 1**

*The adjustment coefficient R when the values of safety loading are 0.1, 0.2 and 0.3*

| R when $\theta = 0.1$ | R when $\theta = 0.2$ | R when $\theta = 0.3$ |
|-----------------------|-----------------------|-----------------------|
| 0.00000002387969      | 0.00000004215254      | 0.00000005754026      |

The result from table 1 showed that the adjustment coefficient increases as the safety loading increases. The table below shows the Tijim's ruin approximation to ruin probability  $\Psi_T(u)$  for initial capital when the values of premium loading  $\theta$  are 0.1, 0.2 and 0.3 respectively

**Table 2**

*Tijim's ruin approximation*

| INITIAL CAPITAL ( $u$ ) | $\Psi_p(u); \theta = 0.1$ | $\Psi_p(u); \theta = 0.2$ | $\Psi_p(u); \theta = 0.3$ |
|-------------------------|---------------------------|---------------------------|---------------------------|
| 11000000                | 0.68826922                | 0.5044192                 | 0.38691274                |
| 11100000                | 0.68652444                | 0.5021012                 | 0.38446642                |
| 11200000                | 0.68478397                | 0.49979345                | 0.38203482                |
| 11300000                | 0.6830478                 | 0.49749588                | 0.37961784                |
| 11400000                | 0.68131592                | 0.49520846                | 0.3772154                 |
| 11500000                | 0.67958833                | 0.49293115                | 0.37482742                |
| 11600000                | 0.677865                  | 0.4906639                 | 0.3724538                 |
| 11700000                | 0.67614594                | 0.48840667                | 0.37009447                |
| 11800000                | 0.67443112                | 0.48615941                | 0.36774934                |
| 11900000                | 0.67272055                | 0.48392209                | 0.36541833                |
| 12000000                | 0.6710142                 | 0.48169465                | 0.36310135                |
| 12100000                | 0.66931207                | 0.47947706                | 0.36079833                |
| 12200000                | 0.66761415                | 0.47726927                | 0.35850917                |
| 12300000                | 0.66592043                | 0.47507124                | 0.3562338                 |
| 12400000                | 0.6642309                 | 0.47288293                | 0.35397214                |
| 12500000                | 0.66254555                | 0.47070429                | 0.3517241                 |
| 12600000                | 0.66086436                | 0.46853529                | 0.34948962                |
| 12700000                | 0.65918733                | 0.46637589                | 0.3472686                 |
| 12800000                | 0.65751445                | 0.46422603                | 0.34506096                |
| 12900000                | 0.6558457                 | 0.46208568                | 0.34286664                |
| 13000000                | 0.65418109                | 0.45995481                | 0.34068554                |
| 13100000                | 0.65252059                | 0.45783336                | 0.3385176                 |
| 13200000                | 0.65086419                | 0.45572129                | 0.33636274                |
| 13300000                | 0.6492119                 | 0.45361858                | 0.33422087                |
| 13400000                | 0.64756369                | 0.45152517                | 0.33209192                |
| 13500000                | 0.64591956                | 0.44944102                | 0.32997582                |
| 13600000                | 0.64427949                | 0.4473661                 | 0.32787249                |
| 13700000                | 0.64264349                | 0.44530036                | 0.32578186                |
| 13800000                | 0.64101153                | 0.44324378                | 0.32370385                |
| 13900000                | 0.63938361                | 0.44119629                | 0.32163838                |
| 14000000                | 0.63775971                | 0.43915788                | 0.31958538                |
| 14100000                | 0.63613984                | 0.43712849                | 0.31754479                |
| 14200000                | 0.63452397                | 0.43510809                | 0.31551652                |
| 14300000                | 0.6329121                 | 0.43309663                | 0.3135005                 |
| 14400000                | 0.63130422                | 0.43109409                | 0.31149667                |
| 14500000                | 0.62970031                | 0.42910042                | 0.30950494                |

|          |            |            |            |
|----------|------------|------------|------------|
| 14600000 | 0.62810038 | 0.42711559 | 0.30752525 |
| 14700000 | 0.62650441 | 0.42513955 | 0.30555753 |
| 14800000 | 0.62491238 | 0.42317227 | 0.30360171 |
| 14900000 | 0.6233243  | 0.4212137  | 0.30165772 |
| 15000000 | 0.62174015 | 0.41926382 | 0.29972548 |
| 15100000 | 0.62015992 | 0.41732258 | 0.29780493 |
| 15200000 | 0.6185836  | 0.41538995 | 0.29589599 |
| 15300000 | 0.61701118 | 0.41346589 | 0.29399861 |
| 15400000 | 0.61544265 | 0.41155036 | 0.29211271 |
| 15500000 | 0.61387801 | 0.40964332 | 0.29023823 |
| 15600000 | 0.61231724 | 0.40774474 | 0.28837509 |
| 15700000 | 0.61076033 | 0.40585459 | 0.28652324 |
| 15800000 | 0.60920728 | 0.40397282 | 0.2846826  |
| 15900000 | 0.60765808 | 0.40209939 | 0.2828531  |
| 16000000 | 0.60611271 | 0.40023428 | 0.28103469 |
| 16100000 | 0.60457116 | 0.39837745 | 0.2792273  |
| 16200000 | 0.60303344 | 0.39652886 | 0.27743085 |
| 16300000 | 0.60149952 | 0.39468847 | 0.2756453  |
| 16400000 | 0.5999694  | 0.39285625 | 0.27387057 |
| 16500000 | 0.59844307 | 0.39103216 | 0.2721066  |
| 16600000 | 0.59692051 | 0.38921618 | 0.27035332 |
| 16700000 | 0.59540173 | 0.38740825 | 0.26861068 |
| 16800000 | 0.59388672 | 0.38560836 | 0.26687861 |
| 16900000 | 0.59237545 | 0.38381646 | 0.26515704 |
| 17000000 | 0.59086793 | 0.38203251 | 0.26344593 |
| 17100000 | 0.58936414 | 0.3802565  | 0.26174519 |
| 17200000 | 0.58786407 | 0.37848837 | 0.26005478 |
| 17300000 | 0.58636773 | 0.3767281  | 0.25837463 |
| 17400000 | 0.58487509 | 0.37497565 | 0.25670468 |
| 17500000 | 0.58338614 | 0.37323098 | 0.25504488 |
| 17600000 | 0.58190089 | 0.37149407 | 0.25339515 |
| 17700000 | 0.58041931 | 0.36976489 | 0.25175545 |
| 17800000 | 0.57894141 | 0.36804338 | 0.25012571 |
| 17900000 | 0.57746717 | 0.36632954 | 0.24850587 |
| 18000000 | 0.57599658 | 0.36462331 | 0.24689588 |
| 18100000 | 0.57452963 | 0.36292467 | 0.24529568 |
| 18200000 | 0.57306632 | 0.36123358 | 0.24370521 |
| 18300000 | 0.57160664 | 0.35955002 | 0.24212441 |
| 18400000 | 0.57015058 | 0.35787394 | 0.24055322 |
| 18500000 | 0.56869812 | 0.35620532 | 0.23899159 |
| 18600000 | 0.56724926 | 0.35454413 | 0.23743946 |
| 18700000 | 0.56580399 | 0.35289032 | 0.23589678 |
| 18800000 | 0.56436231 | 0.35124388 | 0.23436349 |
| 18900000 | 0.5629242  | 0.34960476 | 0.23283954 |
| 19000000 | 0.56148965 | 0.34797294 | 0.23132486 |
| 19100000 | 0.56005866 | 0.34634838 | 0.2298194  |
| 19200000 | 0.55863122 | 0.34473106 | 0.22832312 |
| 19300000 | 0.55720732 | 0.34312094 | 0.22683595 |
| 19400000 | 0.55578694 | 0.34151798 | 0.22535785 |
| 19500000 | 0.55437009 | 0.33992217 | 0.22388875 |
| 19600000 | 0.55295675 | 0.33833346 | 0.2224286  |
| 19700000 | 0.55154692 | 0.33675183 | 0.22097736 |
| 19800000 | 0.55014058 | 0.33517725 | 0.21953497 |
| 19900000 | 0.54873772 | 0.33360968 | 0.21810137 |
| 20000000 | 0.54733835 | 0.3320491  | 0.21667652 |
| 20100000 | 0.54594245 | 0.33049547 | 0.21526037 |
| 20200000 | 0.54455001 | 0.32894877 | 0.21385285 |
| 20300000 | 0.54316102 | 0.32740896 | 0.21245393 |
| 20400000 | 0.54177548 | 0.32587601 | 0.21106354 |
| 20500000 | 0.54039337 | 0.3243499  | 0.20968164 |

|          |            |            |            |
|----------|------------|------------|------------|
| 20600000 | 0.53901469 | 0.3228306  | 0.20830819 |
| 20700000 | 0.53763943 | 0.32131807 | 0.20694312 |
| 20800000 | 0.53626758 | 0.31981228 | 0.20558639 |
| 20900000 | 0.53489914 | 0.31831321 | 0.20423795 |
| 21000000 | 0.53353409 | 0.31682083 | 0.20289776 |
| 21100000 | 0.53217243 | 0.31533511 | 0.20156575 |
| 21200000 | 0.53081414 | 0.31385602 | 0.20024189 |
| 21300000 | 0.52945923 | 0.31238353 | 0.19892612 |
| 21400000 | 0.52810767 | 0.31091761 | 0.19761841 |
| 21500000 | 0.52675947 | 0.30945823 | 0.19631869 |
| 21600000 | 0.52541462 | 0.30800536 | 0.19502692 |
| 21700000 | 0.5240731  | 0.30655899 | 0.19374306 |
| 21800000 | 0.52273491 | 0.30511907 | 0.19246706 |
| 21900000 | 0.52140004 | 0.30368558 | 0.19119887 |
| 22000000 | 0.52006849 | 0.30225849 | 0.18993844 |
| 22100000 | 0.51874024 | 0.30083778 | 0.18868574 |
| 22200000 | 0.51741528 | 0.29942342 | 0.18744071 |
| 22300000 | 0.51609362 | 0.29801537 | 0.18620331 |
| 22400000 | 0.51477523 | 0.29661361 | 0.18497349 |
| 22500000 | 0.51346012 | 0.29521812 | 0.18375121 |
| 22600000 | 0.51214827 | 0.29382887 | 0.18253642 |
| 22700000 | 0.51083968 | 0.29244582 | 0.18132908 |
| 22800000 | 0.50953434 | 0.29106896 | 0.18012915 |
| 22900000 | 0.50823224 | 0.28969825 | 0.17893658 |
| 23000000 | 0.50693337 | 0.28833368 | 0.17775133 |
| 23100000 | 0.50563772 | 0.2869752  | 0.17657335 |
| 23200000 | 0.5043453  | 0.2856228  | 0.1754026  |
| 23300000 | 0.50305608 | 0.28427645 | 0.17423903 |
| 23400000 | 0.50177006 | 0.28293612 | 0.17308262 |
| 23500000 | 0.50048724 | 0.28160179 | 0.17193331 |
| 23600000 | 0.4992076  | 0.28027343 | 0.17079105 |
| 23700000 | 0.49793114 | 0.27895101 | 0.16965582 |
| 23800000 | 0.49665785 | 0.27763451 | 0.16852756 |
| 23900000 | 0.49538772 | 0.27632391 | 0.16740624 |
| 24000000 | 0.49412075 | 0.27501917 | 0.16629182 |
| 24100000 | 0.49285692 | 0.27372027 | 0.16518425 |
| 24200000 | 0.49159623 | 0.27242719 | 0.16408349 |
| 24300000 | 0.49033868 | 0.2711399  | 0.1629895  |
| 24400000 | 0.48908424 | 0.26985838 | 0.16190225 |
| 24500000 | 0.48783293 | 0.2685826  | 0.16082169 |
| 24600000 | 0.48658472 | 0.26731253 | 0.15974778 |
| 24700000 | 0.48533961 | 0.26604815 | 0.15868048 |
| 24800000 | 0.4840976  | 0.26478944 | 0.15761976 |
| 24900000 | 0.48285868 | 0.26353637 | 0.15656557 |
| 25000000 | 0.48162283 | 0.26228891 | 0.15551788 |
| 25100000 | 0.48039005 | 0.26104705 | 0.15447664 |
| 25200000 | 0.47916034 | 0.25981076 | 0.15344183 |
| 25300000 | 0.47793368 | 0.25858    | 0.15241339 |
| 25400000 | 0.47671007 | 0.25735477 | 0.1513913  |
| 25500000 | 0.4754895  | 0.25613503 | 0.15037551 |
| 25600000 | 0.47427196 | 0.25492077 | 0.149366   |
| 25700000 | 0.47305745 | 0.25371195 | 0.14836271 |
| 25800000 | 0.47184596 | 0.25250855 | 0.14736561 |
| 25900000 | 0.47063748 | 0.25131056 | 0.14637467 |
| 26000000 | 0.46943201 | 0.25011794 | 0.14538985 |
| 26100000 | 0.46822953 | 0.24893067 | 0.14441112 |
| 26200000 | 0.46703004 | 0.24774874 | 0.14343843 |
| 26300000 | 0.46583353 | 0.2465721  | 0.14247175 |
| 26400000 | 0.46464    | 0.24540076 | 0.14151105 |
| 26500000 | 0.46344944 | 0.24423467 | 0.14055628 |

|          |            |            |            |
|----------|------------|------------|------------|
| 26600000 | 0.46226183 | 0.24307382 | 0.13960743 |
| 26700000 | 0.46107718 | 0.24191818 | 0.13866444 |
| 26800000 | 0.45989547 | 0.24076774 | 0.13772728 |
| 26900000 | 0.4587167  | 0.23962246 | 0.13679593 |
| 27000000 | 0.45754087 | 0.23848233 | 0.13587034 |
| 27100000 | 0.45636795 | 0.23734733 | 0.13495048 |
| 27200000 | 0.45519796 | 0.23621742 | 0.13403632 |
| 27300000 | 0.45403087 | 0.2350926  | 0.13312782 |
| 27400000 | 0.45286669 | 0.23397283 | 0.13222495 |
| 27500000 | 0.4517054  | 0.23285809 | 0.13132768 |
| 27600000 | 0.450547   | 0.23174837 | 0.13043596 |
| 27700000 | 0.44939149 | 0.23064364 | 0.12954978 |
| 27800000 | 0.44823884 | 0.22954388 | 0.12866909 |
| 27900000 | 0.44708907 | 0.22844907 | 0.12779387 |
| 28000000 | 0.44594215 | 0.22735918 | 0.12692407 |
| 28100000 | 0.44479809 | 0.2262742  | 0.12605968 |
| 28200000 | 0.44365688 | 0.2251941  | 0.12520065 |
| 28300000 | 0.4425185  | 0.22411887 | 0.12434695 |
| 28400000 | 0.44138296 | 0.22304847 | 0.12349856 |
| 28500000 | 0.44025024 | 0.22198289 | 0.12265544 |
| 28600000 | 0.43912034 | 0.22092211 | 0.12181756 |
| 28700000 | 0.43799325 | 0.21986611 | 0.12098488 |
| 28800000 | 0.43686897 | 0.21881486 | 0.12015738 |
| 28900000 | 0.43574749 | 0.21776835 | 0.11933503 |
| 29000000 | 0.4346288  | 0.21672656 | 0.11851779 |
| 29100000 | 0.43351289 | 0.21568945 | 0.11770564 |
| 29200000 | 0.43239976 | 0.21465703 | 0.11689854 |
| 29300000 | 0.4312894  | 0.21362925 | 0.11609647 |
| 29400000 | 0.4301818  | 0.21260611 | 0.11529939 |
| 29500000 | 0.42907696 | 0.21158758 | 0.11450728 |
| 29600000 | 0.42797488 | 0.21057365 | 0.1137201  |
| 29700000 | 0.42687553 | 0.20956428 | 0.11293782 |
| 29800000 | 0.42577892 | 0.20855947 | 0.11216042 |
| 29900000 | 0.42468505 | 0.20755919 | 0.11138787 |
| 30000000 | 0.42359389 | 0.20656343 | 0.11062014 |

### Ruin Probability Using De-Vylder's Approximation Based on the Values of the Adjustment Coefficient $R$

The De-Vylder's approximation to ruin probability  $\Psi_D(u)$  for initial capital  $u$  when the values of premium loading  $\theta$  are 0.1, 0.2 and 0.3 respectively.

**Table 3**

*The De-Vylder's approximation to ruin probability*

| INITIAL CAPITAL ( $u$ ) | $\psi_u(u); \theta = 0.1$ | $\psi_u(u); \theta = 0.2$ | $\psi_u(u); \theta = 0.3$ |
|-------------------------|---------------------------|---------------------------|---------------------------|
| 11000000                | 0.66905584                | 0.48013436                | 0.36286841                |
| 11100000                | 0.66734262                | 0.47792727                | 0.36059877                |
| 11200000                | 0.6656338                 | 0.47573032                | 0.35834332                |
| 11300000                | 0.66392935                | 0.47354346                | 0.35610199                |
| 11400000                | 0.66222927                | 0.47136666                | 0.35387467                |
| 11500000                | 0.66053354                | 0.46919987                | 0.35166128                |
| 11600000                | 0.65884215                | 0.46704304                | 0.34946173                |
| 11700000                | 0.65715509                | 0.46489612                | 0.34727595                |
| 11800000                | 0.65547235                | 0.46275907                | 0.34510383                |
| 11900000                | 0.65379393                | 0.46063185                | 0.3429453                 |
| 12000000                | 0.65211979                | 0.4585144                 | 0.34080027                |
| 12100000                | 0.65044995                | 0.45640668                | 0.33866866                |
| 12200000                | 0.64878438                | 0.45430866                | 0.33655038                |
| 12300000                | 0.64712308                | 0.4522028                 | 0.33444535                |
| 12400000                | 0.64546603                | 0.4501415                 | 0.33235349                |
| 12500000                | 0.64381322                | 0.44807227                | 0.33027471                |

|          |            |            |            |
|----------|------------|------------|------------|
| 12600000 | 0.64216465 | 0.44601256 | 0.32820893 |
| 12700000 | 0.6405203  | 0.44396232 | 0.32615608 |
| 12800000 | 0.63888016 | 0.4419215  | 0.32411606 |
| 12900000 | 0.63724422 | 0.43989006 | 0.3220888  |
| 13000000 | 0.63561246 | 0.43786796 | 0.32007423 |
| 13100000 | 0.63398489 | 0.43585515 | 0.31807225 |
| 13200000 | 0.63236148 | 0.4338516  | 0.3160828  |
| 13300000 | 0.63074223 | 0.43185726 | 0.31410578 |
| 13400000 | 0.62912713 | 0.42987208 | 0.31214114 |
| 13500000 | 0.62751616 | 0.42789603 | 0.31018878 |
| 13600000 | 0.62590932 | 0.42592907 | 0.30824864 |
| 13700000 | 0.62430659 | 0.42397114 | 0.30632063 |
| 13800000 | 0.62270796 | 0.42202222 | 0.30440468 |
| 13900000 | 0.62111343 | 0.42008225 | 0.30250071 |
| 14000000 | 0.61952299 | 0.41815121 | 0.30060865 |
| 14100000 | 0.61793661 | 0.41622904 | 0.29872843 |
| 14200000 | 0.6163543  | 0.4143157  | 0.29685996 |
| 14300000 | 0.61477604 | 0.41241116 | 0.29500318 |
| 14400000 | 0.61320182 | 0.41051538 | 0.29315802 |
| 14500000 | 0.61163163 | 0.40862831 | 0.2913244  |
| 14600000 | 0.61006546 | 0.40674991 | 0.28950224 |
| 14700000 | 0.6085033  | 0.40488015 | 0.28769149 |
| 14800000 | 0.60694514 | 0.40301898 | 0.28589206 |
| 14900000 | 0.60539097 | 0.40116637 | 0.28410388 |
| 15000000 | 0.60384079 | 0.39932228 | 0.28232689 |
| 15100000 | 0.60229457 | 0.39748666 | 0.28056101 |
| 15200000 | 0.60075231 | 0.39565948 | 0.27880618 |
| 15300000 | 0.599214   | 0.3938407  | 0.27706232 |
| 15400000 | 0.59767963 | 0.39203028 | 0.27532938 |
| 15500000 | 0.59614918 | 0.39022819 | 0.27360727 |
| 15600000 | 0.59462266 | 0.38843437 | 0.27189593 |
| 15700000 | 0.59310005 | 0.38664881 | 0.27019529 |
| 15800000 | 0.59158133 | 0.38487145 | 0.2685053  |
| 15900000 | 0.5900665  | 0.38310226 | 0.26682587 |
| 16000000 | 0.58855556 | 0.3813412  | 0.26515695 |
| 16100000 | 0.58704848 | 0.37958824 | 0.26349846 |
| 16200000 | 0.58554526 | 0.37784334 | 0.26185035 |
| 16300000 | 0.58404589 | 0.37610645 | 0.26021255 |
| 16400000 | 0.58255036 | 0.37437756 | 0.25858499 |
| 16500000 | 0.58105866 | 0.37265661 | 0.25696762 |
| 16600000 | 0.57957077 | 0.37094357 | 0.25536035 |
| 16700000 | 0.5780867  | 0.3692384  | 0.25376315 |
| 16800000 | 0.57660643 | 0.36754107 | 0.25217593 |
| 16900000 | 0.57512995 | 0.36585155 | 0.25059864 |
| 17000000 | 0.57365725 | 0.36416979 | 0.24903121 |
| 17100000 | 0.57218832 | 0.36249577 | 0.24747359 |
| 17200000 | 0.57072315 | 0.36082943 | 0.24592571 |
| 17300000 | 0.56926173 | 0.35917076 | 0.24438751 |
| 17400000 | 0.56780406 | 0.35751971 | 0.24285894 |
| 17500000 | 0.56635012 | 0.35587626 | 0.24133992 |
| 17600000 | 0.5648999  | 0.35424035 | 0.23983041 |
| 17700000 | 0.56345339 | 0.35261197 | 0.23833033 |
| 17800000 | 0.56201059 | 0.35099107 | 0.23683964 |
| 17900000 | 0.56057149 | 0.34937762 | 0.23535828 |
| 18000000 | 0.55913606 | 0.34777159 | 0.23388617 |
| 18100000 | 0.55770432 | 0.34617295 | 0.23242328 |
| 18200000 | 0.55627624 | 0.34458165 | 0.23096954 |
| 18300000 | 0.55485182 | 0.34299766 | 0.22952489 |
| 18400000 | 0.55343104 | 0.34142096 | 0.22808927 |
| 18500000 | 0.5520139  | 0.33985151 | 0.22666264 |

|          |            |            |            |
|----------|------------|------------|------------|
| 18600000 | 0.55060039 | 0.33828927 | 0.22524492 |
| 18700000 | 0.5491905  | 0.33673421 | 0.22383608 |
| 18800000 | 0.54778423 | 0.3351863  | 0.22243605 |
| 18900000 | 0.54638155 | 0.3336455  | 0.22104477 |
| 19000000 | 0.54498246 | 0.33211179 | 0.21966219 |
| 19100000 | 0.54358696 | 0.33058513 | 0.21828827 |
| 19200000 | 0.54219503 | 0.32906549 | 0.21692294 |
| 19300000 | 0.54080666 | 0.32755283 | 0.21556614 |
| 19400000 | 0.53942185 | 0.32604712 | 0.21421784 |
| 19500000 | 0.53804058 | 0.32454834 | 0.21287796 |
| 19600000 | 0.53666286 | 0.32305644 | 0.21154647 |
| 19700000 | 0.53528866 | 0.32157141 | 0.2102233  |
| 19800000 | 0.53391797 | 0.3200932  | 0.20890841 |
| 19900000 | 0.5325508  | 0.31862178 | 0.20760175 |
| 20000000 | 0.53118713 | 0.31715713 | 0.20630326 |
| 20100000 | 0.52982695 | 0.31569922 | 0.20501289 |
| 20200000 | 0.52847026 | 0.314248   | 0.20373059 |
| 20300000 | 0.52711704 | 0.31280345 | 0.20245631 |
| 20400000 | 0.52576728 | 0.31136555 | 0.20119    |
| 20500000 | 0.52442098 | 0.30993425 | 0.19993161 |
| 20600000 | 0.52307812 | 0.30850954 | 0.1986811  |
| 20700000 | 0.52173871 | 0.30709137 | 0.1974384  |
| 20800000 | 0.52040273 | 0.30567972 | 0.19620348 |
| 20900000 | 0.51907016 | 0.30427457 | 0.19497628 |
| 21000000 | 0.51774101 | 0.30287587 | 0.19375676 |
| 21100000 | 0.51641526 | 0.3014836  | 0.19254486 |
| 21200000 | 0.51509291 | 0.30009773 | 0.19134055 |
| 21300000 | 0.51377394 | 0.29871823 | 0.19014376 |
| 21400000 | 0.51245835 | 0.29734507 | 0.18895447 |
| 21500000 | 0.51114613 | 0.29597823 | 0.18777261 |
| 21600000 | 0.50983727 | 0.29461767 | 0.18659814 |
| 21700000 | 0.50853176 | 0.29326336 | 0.18543102 |
| 21800000 | 0.50722959 | 0.29191528 | 0.1842712  |
| 21900000 | 0.50593076 | 0.29057339 | 0.18311864 |
| 22000000 | 0.50463525 | 0.28923767 | 0.18197328 |
| 22100000 | 0.50334306 | 0.2879081  | 0.18083509 |
| 22200000 | 0.50205418 | 0.28658463 | 0.17970401 |
| 22300000 | 0.5007686  | 0.28526725 | 0.17858002 |
| 22400000 | 0.49948632 | 0.28395592 | 0.17746305 |
| 22500000 | 0.49820731 | 0.28265063 | 0.17635306 |
| 22600000 | 0.49693158 | 0.28135133 | 0.17525002 |
| 22700000 | 0.49565912 | 0.28005801 | 0.17415388 |
| 22800000 | 0.49438991 | 0.27877063 | 0.1730646  |
| 22900000 | 0.49312396 | 0.27748916 | 0.17198213 |
| 23000000 | 0.49186125 | 0.27621359 | 0.17090643 |
| 23100000 | 0.49060177 | 0.27494389 | 0.16983745 |
| 23200000 | 0.48934551 | 0.27368002 | 0.16877517 |
| 23300000 | 0.48809248 | 0.27242195 | 0.16771953 |
| 23400000 | 0.48684265 | 0.27116968 | 0.16667049 |
| 23500000 | 0.48559602 | 0.26992316 | 0.16562801 |
| 23600000 | 0.48435258 | 0.26868236 | 0.16459205 |
| 23700000 | 0.48311233 | 0.26744728 | 0.16356257 |
| 23800000 | 0.48187525 | 0.26621787 | 0.16253953 |
| 23900000 | 0.48064134 | 0.26499411 | 0.16152289 |
| 24000000 | 0.47941059 | 0.26377597 | 0.16051261 |
| 24100000 | 0.478183   | 0.26256344 | 0.15950865 |
| 24200000 | 0.47695854 | 0.26135648 | 0.15851097 |
| 24300000 | 0.47573722 | 0.26015507 | 0.15751953 |
| 24400000 | 0.47451903 | 0.25895918 | 0.15653428 |
| 24500000 | 0.47330396 | 0.25776879 | 0.15555521 |

|          |            |            |            |
|----------|------------|------------|------------|
| 24600000 | 0.472092   | 0.25658387 | 0.15458225 |
| 24700000 | 0.47088314 | 0.2554044  | 0.15361538 |
| 24800000 | 0.46967738 | 0.25423034 | 0.15265456 |
| 24900000 | 0.4684747  | 0.25306169 | 0.15169975 |
| 25000000 | 0.46727511 | 0.25189841 | 0.15075091 |
| 25100000 | 0.46607859 | 0.25074047 | 0.149808   |
| 25200000 | 0.46488513 | 0.24958786 | 0.14887099 |
| 25300000 | 0.46369472 | 0.24844055 | 0.14793985 |
| 25400000 | 0.46250737 | 0.24729851 | 0.14701452 |
| 25500000 | 0.46132305 | 0.24616172 | 0.14609499 |
| 25600000 | 0.46014177 | 0.24503016 | 0.1451812  |
| 25700000 | 0.45896351 | 0.24390379 | 0.14427314 |
| 25800000 | 0.45778827 | 0.24278261 | 0.14337075 |
| 25900000 | 0.45661604 | 0.24166658 | 0.142474   |
| 26000000 | 0.45544681 | 0.24055568 | 0.14158287 |
| 26100000 | 0.45428058 | 0.23944988 | 0.14069731 |
| 26200000 | 0.45311733 | 0.23834917 | 0.13981728 |
| 26300000 | 0.45195706 | 0.23725352 | 0.13894276 |
| 26400000 | 0.45079976 | 0.23616291 | 0.13807372 |
| 26500000 | 0.44964542 | 0.23507731 | 0.1372101  |
| 26600000 | 0.44849404 | 0.2339967  | 0.13635189 |
| 26700000 | 0.44734561 | 0.23292105 | 0.13549905 |
| 26800000 | 0.44620012 | 0.23185035 | 0.13465154 |
| 26900000 | 0.44505756 | 0.23078458 | 0.13380933 |
| 27000000 | 0.44391793 | 0.2297237  | 0.13297239 |
| 27100000 | 0.44278122 | 0.2286677  | 0.13214068 |
| 27200000 | 0.44164741 | 0.22761655 | 0.13131418 |
| 27300000 | 0.44051651 | 0.22657024 | 0.13049285 |
| 27400000 | 0.43938851 | 0.22552873 | 0.12967665 |
| 27500000 | 0.43826339 | 0.22449201 | 0.12886556 |
| 27600000 | 0.43714116 | 0.22346006 | 0.12805954 |
| 27700000 | 0.4360218  | 0.22243285 | 0.12725856 |
| 27800000 | 0.4349053  | 0.22141037 | 0.12646259 |
| 27900000 | 0.43379167 | 0.22039258 | 0.12567161 |
| 28000000 | 0.43268088 | 0.21937947 | 0.12488556 |
| 28100000 | 0.43157294 | 0.21837102 | 0.12410444 |
| 28200000 | 0.43046784 | 0.21736721 | 0.1233282  |
| 28300000 | 0.42936557 | 0.21636801 | 0.12255682 |
| 28400000 | 0.42826612 | 0.2153734  | 0.12179026 |
| 28500000 | 0.42716948 | 0.21438336 | 0.12102849 |
| 28600000 | 0.42607566 | 0.21339788 | 0.12027149 |
| 28700000 | 0.42498463 | 0.21241693 | 0.11951923 |
| 28800000 | 0.4238964  | 0.21144048 | 0.11877167 |
| 28900000 | 0.42281095 | 0.21046853 | 0.11802878 |
| 29000000 | 0.42172828 | 0.20950104 | 0.11729055 |
| 29100000 | 0.42064839 | 0.208538   | 0.11655693 |
| 29200000 | 0.41957126 | 0.20757938 | 0.1158279  |
| 29300000 | 0.41849689 | 0.20662517 | 0.11510342 |
| 29400000 | 0.41742527 | 0.20567535 | 0.11438348 |
| 29500000 | 0.41635639 | 0.2047299  | 0.11366805 |
| 29600000 | 0.41529026 | 0.20378879 | 0.11295708 |
| 29700000 | 0.41422685 | 0.20285201 | 0.11225057 |
| 29800000 | 0.41316616 | 0.20191953 | 0.11154847 |
| 29900000 | 0.41210819 | 0.20099134 | 0.11085077 |
| 30000000 | 0.41105293 | 0.20006742 | 0.11015743 |

### Results and Discussion

We recall from equation (43) that  $E(U(\xi)) = u + \pi\xi - \mu\lambda\xi$

But by the Waald's identity,

$$\lim_{\xi \rightarrow \infty} \left\{ \frac{\mathbf{E}(U(\xi))}{\xi} \right\} = \lim_{\xi \rightarrow \infty} \left( \frac{u + \pi\xi - \mu\lambda\xi}{\xi} \right) = \lim_{\xi \rightarrow \infty} \left( \frac{u}{\xi} + \frac{\pi\xi - \mu\lambda\xi}{\xi} \right) \quad (144a)$$

$$\lim_{\xi \rightarrow \infty} \left\{ \frac{\mathbf{E}(U(\xi))}{\xi} \right\} = \pi - \mu\lambda \quad (144b)$$

By the reason of the strong law of large numbers, it is clear that in the long run  $U(\xi)$  will converge to  $-\infty$  (a.s) when  $\pi > \mu\lambda$  whereas if  $\pi < \mu\lambda$  then  $U(\xi)$  will converge to  $\infty$  (a.s) for all  $\xi$ . However, if  $\pi = \mu\lambda$ , then  $\liminf_{\xi \rightarrow \infty} \{U(\xi)\} = \limsup_{\xi \rightarrow \infty} \{U(\xi)\} = -\infty$ .

The result from table 2 and 3 suggests that the probability of ruin decreases as the initial capital increases signifying that the higher the initial capital  $u$  the higher the survival probability of the model. It also shows that the probability of ruin decreases as the safety loading  $\theta$  increases indicating that the bigger the adjustment coefficient  $R$  the higher survival probability the model has. Table 1 shows the values of the adjustment coefficient  $R$  for the corresponding values of the safety loadings 0.1, 0.2 and 0.3 respectively. The adjustment coefficient therefore increases as the safety loading increases. The result from tables 2-3 showed that as the level of ruin probabilities decreases, the size of the initial capital increases establishing an inverse linear relationship and consequently. Thus a straight line could be applied to forecast future initial capital barring core changes in the risk parameters such as the safety loading and claim sizes. In the event of high level of ruin probabilities, then there will be an anticipated bigger risk appetite within the insurer's underwriting profile implying a small provision for contingency regarding insolvencies such that where the anticipated ruin probabilities declines, there will be a small risk perception adopted and therefore a bigger provision for the anticipated ruin. The stress analysis conducted in the Tables 2-3 reveals that the probability of ruin decreases as the safety loading  $\theta$  increases implying that the higher the adjustment coefficient  $R$ , the higher the survival probability the models have. Computational evidence from tables 2-3 shows that the Tijim's approximation to ruin probability is higher than the De-Vylder's approximation at the same level of initial capital and safety loading. The implication is that the De-Vylder's approximation to ruin probabilities is an improvement over Tijim's ruin model and hence De-Vylder's approximation is recommended for the insurance firm. From the foregoing, the results obtained could be employed to advise the insurance firms through the regulatory authorities to enshrine policy framework which can forestall pervasive consequences of ruin and consequently, the regulatory authorities should therefore enforce policy recommendations on improved minimum capital to escape ruinous conditions.

### Conclusion

From our discussions, it is necessary for an insurance firm to conduct underwriting business above a defined level of income assumed set above zero or a specified threshold. The time that ruin occurs is then  $\xi = \inf \{t > 0 | U_t \leq 0\}$  and consequently  $\psi(u) = \mathbf{P}(\xi < \infty | U_0 = u)$ . Therefore, the continuous time minimum initial capital requires that the surplus be closely examined to ascertain that ruin does not occur. In our computations, the performance of the approximation was checked by comparing the Tijim's and De-Vylder's ruin probabilities using real claims data from a Nigerian insurance company. From the computed results, a high adjustment coefficient reduces the ruin probability of an underwriter whereas a high initial capital increases the solvency of an insurance company. Verified results from our estimations, De-Vylder's ruin probabilities seems more reliable as the fundamental principle is to replace surplus process with the one characterized by exponentially distributed losses and such that the first three moments coincide. Consequently, the ruin probability using De-Vylder's model is less than that of the Tijim's approximation at the same level of initial capital and safety loading. A sufficient minimum insurance capital should be set by the management of an insurance company in order to ensure the solvency of the company. Again an adequate

safety loading should also be advised such that the company will not enter into ruin in a foreseeable long run.

### Acknowledgement

The authors acknowledge the inputs of FarWestern University and the Reviewers for conducting this research.

### Conflict of Interest

The author(s) declared no potential conflict of interest with respect to the research, authorship and/or publication of this article.

### References

- Burnecki K, Mista P and Weron A. (2003). A new De Vylder type approximation of the ruin probability in infinite time. *Research Report HSC/03/5*. Hugo Steinhaus Center for Stochastic Methods, Institute of Mathematics Wroclaw University of Technology.
- Burnecki, Teuerle, Wilkowska (2019). De Vylder type approximation of the ruin probability for the insurer-reinsurer model. *Journal of Mathematical application Vol. 47(1)*, 5-24.
- Centeno M. L. (1986). Measuring the effects of reinsurance by adjustment coefficient. *Journal of Insurance, Mathematics and Economics*. 5(2), 169-182.
- Cheng, J., Gao Y., and Wang D. (2016). Ruin probabilities for a perturbed risk model with stochastic premiums and constant interest force. *Journal Of Inequalities and Applications*. doi: 10.1186/s13660-016-1135-8
- Guerra M. and Centeno M. L (2008). Optimal reinsurance policy: The adjustment coefficient and the expected utility criteria. *Insurance: Mathematics and Economics*, 42(2), 529-539.
- Huang Y, Li J, Liu H and Yu W. (2021). Estimating ruin probability in an insurance risk model with stochastic premium income based on the CFS method. *Journal of Mathematics, Economics and Insurance* 2021, 9(9) 982; Doi 10.3390/math9090982
- Karageyik B.B and Sahin S. (2016). A review on optimal reinsurance under ruin probability constraint. *Journal of Statisticians and Actuarial Sciences*, IDIA 9(1), 26-36
- Korzeniowski A (2023). Ruin probability for risk model with random premiums. *Journal of Mathematical Finance*, 13(2), 171-179. DOI: 10.4236/jmf.2023.132011
- Liang Z. and Guo J. (2007). Optimal proportional reinsurance and ruin probability. *Stochastic Models*, 23(2): 333-350. DOI: .org/10.1080/15326340701300894
- Luesamai A. (2021). Lower and Upper bounds of the ultimate ruin probability in a discrete time risk model with proportional reinsurance and investment. *Journal of Risk management and Insurance*. 25(1), 595-614
- Michna Z. (2020). Ruin probabilities for two collaborating insurance companies. *Journal of Probability and Mathematical Statistics*. 40(2), 369-386.
- Ogungbenle G.M. (2024). Threshold net profit condition in predicting the insurer's probability of ruin. *South Asian Journal of Finance*, 4(1), 66-85.
- Santanaa, D. J. and Rincon L. (2023). Ruin probabilities as functions of the roots of a polynomial. *Modern Stochastics: Theory and Applications* 10(3), 1–20 <https://doi.org/10.15559/23-VMSTA226>