

A polynomial expansion to approximate the ultimate ruin probability in the Cramer-Lundberg ruin model

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Outlines

- 1 Introduction
- 2 Natural Exponential Family with Quadratic Variance Function (NEF-QVF)
- 3 The Cramer-Lundberg Ruin Model
- 4 Simulation Results

Executive summary

Main Goal

Work out a new numerical method to approximate ruin probabilities

Main Idea

Polynomial expansion of a function through orthogonal projection

- ↪ Change of measure via Natural Exponential Family with Quadratic Variance Function
- ↪ Construction of a polynomial orthogonal system w.r.t this probability measure

Achievement

Approximation of the ultimate ruin probability in the Cramer-Lundberg ruin model with phase type claim size

Notations

dF is an univariate Probability Measure (UPM). Denote by :

- F its Cumulated Distribution Function (CDF).
- $f = F'$ its Probability Density Function (PDF) w.r.t. to a positive measure.
- $\widehat{F}(\theta) = \int e^{\theta x} dF(x)$ its Moment Generating Function (MGF).
- $\kappa(\theta) = \ln(\widehat{F}(\theta))$ its Cumulant Generating Function (CGF).

Denote by $L^2(dF)$ the function space such that :

- If $f \in L^2(dF)$ then $\int f^2(x) dF(x) < \infty$.

$L^2(F)$ is a normed vector space with :

$$\|f\|^2 = \langle f, f \rangle = \int f^2(x) dF(x).$$

Natural Exponential Family

Let F be an univariate CDF possessing a MGF in a neighborhood of 0.

- $\{F_\theta; \theta \in \Theta\}$ is the NEF generated by F , with :

$$dF_\theta(x) = \exp(\theta x - \kappa(\theta))dF(x),$$

and Θ the definition domain of κ .

- The mean of F_θ is

$$\mu = \int x dF_\theta(x) = \kappa'(\theta),$$

where

- $\kappa' : \theta \rightarrow \kappa'(\theta)$ is a bijection.
 $\hookrightarrow \phi : \mu \rightarrow \phi(\mu) = \kappa'^{-1}(\mu) = \theta$ its inverse function.
- $\{F_\mu; \mu \in \kappa'(\Theta) = M\}$ is an equivalent definition for NEF with :

$$dF_\mu(x) = \exp(\phi(\mu)x - \kappa(\phi(\mu)))dF(x).$$

Quadratic Variance Function

The Variance Function (VF) of a NEF F_μ is :

$$V(\mu) = \int (x - \mu)^2 dF_\mu(x) = \kappa''(\mu).$$

The VF is said Quadratic if :

$$V(\mu) = a + b\mu + c\mu^2.$$

The NEF-QVF contains six types of distributions :

- Normal
- Gamma
- Hyperbolic
- Binomial
- Negative Binomial
- Poisson

Orthogonal polynomials for NEF-QVF Distributions

Define by $\{F_\mu; \mu \in M\}$ a NEF-QVF generated by F with mean μ_0 .

- $f(x, \mu)$ a NEF-QVF PDF proportional to $\exp(\phi(\mu)x - \kappa(\phi(\mu)))$ w.r.t. F :

$$Q_n(x, \mu) = V^n(\mu) \left\{ \frac{\partial^n}{\partial \mu^n} f(x, \mu) \right\} / f(x, \mu),$$

is a polynomial of degree n both in μ and x .

- Note that $f(x, \mu_0) = 1$. Then

$$Q_n(x) = Q_n(x, \mu_0) = V^n(\mu_0) \left\{ \frac{\partial^n}{\partial \mu^n} f(x, \mu) \right\}_{\mu=\mu_0}.$$

$\{Q_n\}$ is an orthogonal polynomials system such that :

$$\langle Q_n(x), Q_m(x) \rangle = \int Q_n(x) Q_m(x) dF(x) = \|Q_n\|^2 \delta_{nm}.$$

For a full description of the NEF-QVF see Morris [1982] [1].

Polynomial Expansion and Truncations

- The polynomials are dense in $L^2(F)$.
 $\hookrightarrow \{Q_n\}$ is therefore an orthogonal basis of $L^2(F)$.
- dF_X a probability measure associated to some random variable X .
 $\hookrightarrow \frac{dF_X}{dF}$ PDF w.r.t. dF
- If $\frac{dF_X}{dF} \in L^2(F)$ we have :

$$\frac{dF_X}{dF}(x) = \sum_{n \in \mathbb{N}} \left\langle \frac{dF_X}{dF}, \frac{Q_n}{\|Q_n\|} \right\rangle \frac{Q_n(x)}{\|Q_n\|} = \sum_{n \in \mathbb{N}} E(Q_n(X)) \frac{Q_n(x)}{\|Q_n\|^2}.$$

- The CDF F_X is then :

$$F_X(x) = \sum_{n \in \mathbb{N}} E(Q_n(X)) \frac{\int_{-\infty}^x Q_n(y) dF(y)}{\|Q_n\|^2}.$$

Approximations are then obtained by truncation :

$$F_X^K(x) = \sum_{n=0}^K E(Q_n(X)) \frac{\int_{-\infty}^x Q_n(y) dF(y)}{\|Q_n\|^2}.$$

Definition and hypothesis

Denote by $\{R(t); t \geq 0\}$ the Risk Reserve Process :

$$R(t) = u + pt - \sum_{i=1}^{N(t)} U_i,$$

where

- u is the initial reserve.
- p is the constant premium rate per unit of time.
- $N(t)$ is a Poisson process with intensity β .
- $\{U_i\}_{i \in \mathbb{N}^*}$ are **i.i.d.** non-negative random variables with CDF F_U and mean μ .

Let $\{S(t); t \geq 0\}$ be the Surplus process :

$$S(t) = u - R(t).$$

$\eta > 0$ is the safety loading such that :

$$p = (1 + \eta)\beta\mu.$$

Ultimate ruin probability

We denote by $M = \sup\{S(t); t > 0\}$ and we define the ultimate ruin probability by :

$$\psi(u) = P(M > u) = \overline{F_M}(u).$$

Theorem : Pollaczek-Khinchine formula

In the Cramer-Lundberg model, the ruin probability can be written as :

$$\psi(u) = (1 - \rho) \sum_{n=0}^{+\infty} \rho^n \overline{F_{U^I}^{*n}}(u),$$

$$M \stackrel{D}{=} \sum_{i=1}^N U_i^I,$$

where N is geometric with parameter $\rho = \frac{\beta\mu}{p} < 1$ and $F_{U^I}^{*n}$ denotes the n th convolution of F_{U^I} .

See *Ruin probabilities* by Asmussen and Albrecher [2001] [2].

Phase Type Distribution for Claim amounts

- $\{J(t)\}$ is a continuous time Markov process with state space $E = \{1, \dots, d\}$ and generator \mathbf{T} .
- $\{\tilde{J}(t)\}$ is a terminating Markov process with state space $E_\Delta = \{E \cup \Delta\}$. Δ is an absorbing state and the Markov process' generator is :

$$\left(\begin{array}{c|c} \mathbf{T} & \mathbf{t} \\ \hline 0 & 0 \end{array} \right).$$

F_U is of phase-type with representation (α, \mathbf{T}, E) if :

$$F_U(x) = P_\alpha(\zeta \leq x) = 1 - \alpha e^{\mathbf{T}x} \mathbf{e},$$

where

- $\zeta = \inf\{t; \tilde{J}(t) = \Delta\}$.
- α is a vector of size d such as $\alpha_j = P(J(0) = j)$.
- \mathbf{e} is a vector of size d with all components equal to 1.

Ultimate ruin probability with phase claim size

Theorem : An exponential matrix expression

Assume that F_U is of phase-type with representation (α, \mathbf{T}) then :

- \mathbf{M} is zero-modified phase-type with representation $(\alpha_+, \mathbf{T} + t\alpha_+)$, where $\alpha_+ = -\beta.\alpha.\mathbf{T}^{-1}$.
- $\psi(u) = \alpha_+.e^{(\mathbf{T} + \mathbf{t}.\alpha_+)u}.\mathbf{e}$.

In this case the ruin probability is asymptotically exponential :

$$\psi(u) \sim Ce^{-\lambda u},$$

with $-\lambda = \max\{Sp(\mathbf{T} + \mathbf{t}.\alpha_+)\}$.

Polynomial Expansion for the ruin probability

Recall that $M = \sum_{i=1}^N U_i^I$. Then :

$$\begin{aligned} dF_M(x) &= (1 - \rho)\delta_0(dx) + (1 - \rho) \sum_{n=1}^{+\infty} \rho^n dF_{U^I}^{*n}(x) \\ &= (1 - \rho)\delta_0(dx) + dG_M(x). \end{aligned}$$

If $\frac{dG_M}{dF} \in L^2(F)$ then :

$$\frac{dG_M}{dF}(x) = \sum_{n \in \mathbb{N}} \left\langle \frac{dG_M}{dF}, \frac{Q_n}{\|Q_n\|} \right\rangle \frac{Q_n(x)}{\|Q_n\|}.$$

From this we deduce a polynomial expansion for the ruin probability :

$$\psi(u) = \sum_{n \in \mathbb{N}} \left\langle \frac{dG_M}{dF}, \frac{Q_n}{\|Q_n\|} \right\rangle \frac{\int_u^{+\infty} Q_n(y) dF(y)}{\|Q_n\|}.$$

Approximation of the ruin probability though truncation of the polynomial expansion

Theorem : Approximation of the ultimate ruin probability

Let $\{F_\mu; \mu \in M\}$ be a NEF-QVF generated by F of mean μ_0 and $f(x, \mu)$ a NEF-QVF PDF proportional to $\exp(\phi(\mu)x - \kappa(\phi(\mu)))$ w.r.t F .

- If $\frac{dG_M}{dF} \in L^2(dF)$ then :

$$\begin{aligned} \psi^K(u) &= \sum_{n=0}^K V_F(\mu_0)^n \left[\frac{\partial^n}{\partial \mu^n} e^{-\kappa(\phi(\mu))} \left(\widehat{G}_M(\phi(\mu)) \right) \right]_{\mu=\mu_0} \\ &\times \frac{\int_u^{+\infty} Q_n(y) dF(y)}{\|Q_n(x)\|^2} \end{aligned}$$

Choice of the NEF-QVF

dG_M is a defective probability measure supported on $[0, +\infty]$.

Among NEF-QVF the only one supported on $[0, +\infty]$ is generated by the exponential distribution. Then :

$$dF(x) = \xi e^{-\xi x} d\lambda(x)$$

The orthogonal polynomials linked to the exponential measure are the Laguerre ones see Szegő [1939] [3]

- Which value for ξ to complete the integrability condition ?
- $\frac{dG_M}{dF}(u) \sim D e^{-u(\lambda-\xi)}$
 $\hookrightarrow \xi < 2\lambda$

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 - $\hookrightarrow \xi < 2\lambda$

Simulation studies setting

For the ruin model we assume that :

- the premium rate p is equal to 1
- the safety loading η is equal to 33% and consequently $\rho = 0.75$
- The distributions of claim amount are of phase type given by \mathbf{T} and α

A graphic visualization is proposed, we plot the quantity :

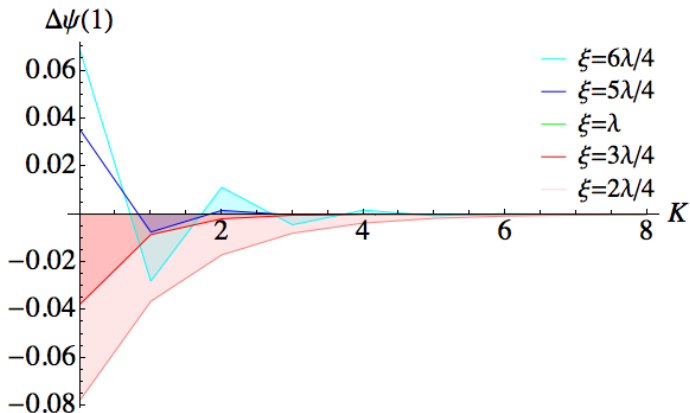
$$\Delta\psi(u) = \psi(u) - \psi^K(u),$$

for an initial reserve u and based on the truncation of order K .

↪ Different values for ξ are tested with one equal to λ .

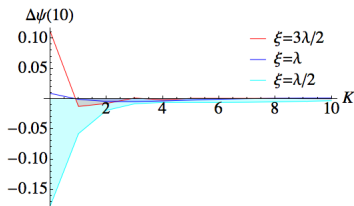
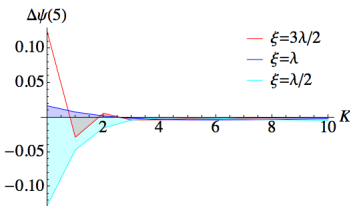
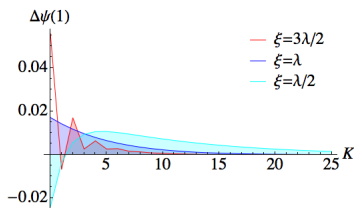
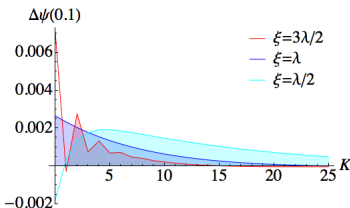
Exponentially distributed claim size

$$f_U(x) = e^{-x} 1_{[0, +\infty)}(x) \quad \lambda = 0,25$$



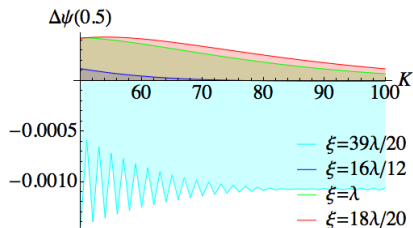
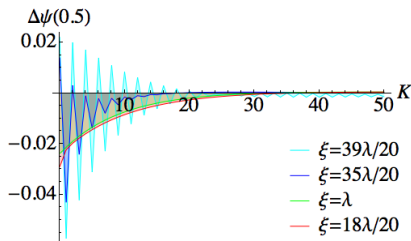
Mixture of Erlang distributed claim size

$$\mathbf{T} = \begin{pmatrix} -1 & 1 & 0 & 0 & 0 \\ 0 & -1 & 1 & 0 & 0 \\ 0 & 0 & -1 & 0 & 0 \\ 0 & 0 & 0 & -\frac{2}{3} & \frac{2}{3} \\ 0 & 0 & 0 & 0 & -\frac{2}{3} \end{pmatrix}, \quad \alpha = \left(\frac{2}{5} \quad 0 \quad 0 \quad \frac{3}{5} \quad 0 \right), \quad \lambda = 0.1205$$



HyperExponentially distributed claim size

$$\mathbf{T} = \begin{pmatrix} -2 & 0 & 0 \\ 0 & -5 & 0 \\ 0 & 0 & -1/2 \end{pmatrix}, \quad \alpha = \begin{pmatrix} \frac{1}{3} & \frac{1}{3} & \frac{1}{3} \end{pmatrix}, \quad \lambda = 0.1495$$



Conclusion

- + A method quite efficient
 - ↪ An approximation as precise as one might want
- + Easy to understand and to implement
- + No discretization of the claim size is needed
- Limited to light tailed distribution

Outlooks :

- Theoretical sensitiveness study of the parameters
- Agregate claim amount distribution
- Finite time ruin probability

References



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Proof for the ruin probability approximation

Let's start from :

$$\begin{aligned}\psi(u) &= \sum_{n \in \mathbb{N}} \left\langle \frac{dG_M}{dF}, Q_n \right\rangle \frac{\int_u^{+\infty} Q_n(y) dF(y)}{\|Q_n\|^2} \\ &= \sum_{n \in \mathbb{N}} \int_{\mathbb{R}} Q_n(x) \frac{dG_M}{dF}(x) dF(x) \frac{\int_u^{+\infty} Q_n(y) dF(y)}{\|Q_n\|^2} \\ &= \sum_{n \in \mathbb{N}} \int_{\mathbb{R}} Q_n(x) dG_M(x) \frac{\int_u^{+\infty} Q_n(y) dF(y)}{\|Q_n\|^2}\end{aligned}\tag{1}$$

recall that :

$$\begin{aligned}Q_n(x) &= V_F(\mu_0)^n \left\{ \frac{\partial^n}{\partial \mu^n} f(x, \mu) \right\}_{\mu=\mu_0} \\ &= V_F(\mu_0)^n \left\{ \frac{\partial^n}{\partial \mu^n} \exp(\phi(\mu)x - \kappa(\phi(\mu))) \right\}_{\mu=\mu_0}\end{aligned}$$

Proof for the ruin probability approximation

Re-injecting in (1), one gets :

$$\begin{aligned}\psi(u) &= \sum_{n \in \mathbb{N}} \int_{\mathbb{R}} V_F(\mu_0)^n \left\{ \frac{\partial^n}{\partial \mu^n} \exp(\phi(\mu)x - \kappa(\phi(\mu))) \right\}_{\mu=\mu_0} dG_M(x) \\ &\quad \times \frac{\int_u^{+\infty} Q_n(y) dF(y)}{\|Q_n\|^2} \\ &= \sum_{n \in \mathbb{N}} V_F(\mu_0)^n \left\{ \frac{\partial^n}{\partial \mu^n} e^{-\kappa(\phi(\mu))} \int_{\mathbb{R}} \exp(\phi(\mu)x) dG_M(x) \right\}_{\mu=\mu_0} \\ &\quad \times \frac{\int_u^{+\infty} Q_n(y) dF(y)}{\|Q_n\|^2} \\ &= \sum_{n \in \mathbb{N}} V_F(\mu_0)^n \left\{ \frac{\partial^n}{\partial \mu^n} e^{-\kappa(\phi(\mu))} \widehat{G}_M(\phi(\mu)) \right\}_{\mu=\mu_0} \frac{\int_u^{+\infty} Q_n(y) dF(y)}{\|Q_n\|^2}\end{aligned}$$