Computation of the Aggregate Claim Amount Distribution Using R and actuar

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Actuarial Risk Modeling Process

- 1 Model costs at the individual level \rightarrow Modeling of loss distributions
- 2 Aggregate risks at the collective level \rightarrow Risk theory
- 3 Determine revenue streams
 - \rightarrow Ratemaking
- 4 Evaluate solvability of insurance portfolio

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What actuar Is

 A package providing additional Actuarial Science functionality to the R statistical system

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- Distributed through the Comprehensive R Archive Network (CRAN)
- Currently provides:
 - 17 additional probability distributions
 - loss modeling facilities
 - aggregate claim amount calculation
 - fitting of credibility models
 - ruin probabilities and related quantities
 - simulation of compound hierarchical models

Yes But...Why R?

Compare

x <- matrix(2, 3, 10:15) vs x ← 2 3ρ9 + 16

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- Multi-platform
- Interactive
- State-of-the-art statistical procedures, random number generators and graphics

Collective Risk Model

Let

- S : aggregate claim amount
- N : number of claims (frequency)
- C_j : amount of claim *j* (severity)

We have the random sum

$$S = C_1 + \cdots + C_N$$

We want to find

$$F_{S}(x) = \Pr[S \le x]$$

= $\sum_{n=0}^{\infty} \Pr[S \le x | N = n] \Pr[N = n]$
= $\sum_{n=0}^{\infty} F_{C}^{*n}(x) \Pr[N = n]$

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What We're Presenting Here Today



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 aggregateDist() \longrightarrow $F_{\mathcal{S}}(x)$

(Insightful user)

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- 2 Use a Normal approximation

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- **1** Carry out the convolutions $F_C^{*k}(x)$ for k = 0, 1, 2, ...
- 2 Use a Normal approximation

$$F_{\mathcal{S}}(\mathbf{x}) \simeq \Phi\left(\frac{\mathbf{x}-\mu_{\mathcal{S}}}{\sigma_{\mathcal{S}}}\right)$$

3 Use the Normal Power II approximation

$$F_{S}(x) \simeq \Phi\left(-\frac{3}{\gamma_{S}} + \sqrt{\frac{9}{\gamma_{S}^{2}} + 1 + \frac{6}{\gamma_{S}}\frac{x - \mu_{S}}{\sigma_{S}}}\right)$$

4 Use simulation:

$$F_{\mathcal{S}}(x) \simeq F_n(x) = \frac{1}{n} \sum_{j=1}^n I\{x_j \le x\}$$

Most Commonly Used Method

5 Recursive method (Panjer algorithm):

$$f_{S}(x) = \frac{1}{1 - af_{C}(0)} \left[(p_{1} - (a + b)p_{0})f_{C}(x) + \sum_{y=1}^{\min(x,m)} (a + by/x)f_{C}(y)f_{S}(x - y) \right]$$

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with

$$f_{\mathcal{S}}(0) = P_{\mathcal{N}}(f_{\mathcal{C}}(0))$$

Most Commonly Used Method

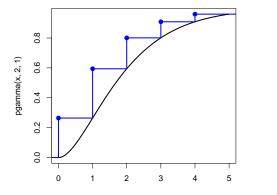
5 Recursive method (Panjer algorithm):

$$f_{S}(x) = \frac{1}{1 - af_{C}(0)} \left[(p_{1} - (a + b)p_{0})f_{C}(x) + \sum_{y=1}^{\min(x,m)} (a + by/x)f_{C}(y)f_{S}(x - y) \right]$$

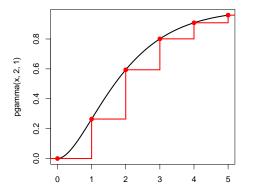
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with

$$f_{\mathcal{S}}(0) = P_{\mathcal{N}}(f_{\mathcal{C}}(0))$$

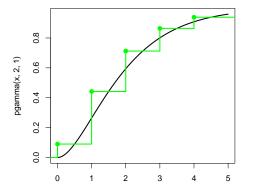


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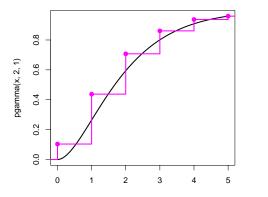


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Computing the Aggregate Claim Amount Distribution

aggregateDist() is the unified interface to all 5 supported methods

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- Computer intensive calculations are done in C
- Output is a **function** to compute $F_S(x)$ in any x
- R methods to plot and compute related quantities

Example

Assume

 $N \sim \text{Poisson}(10)$ $C \sim \text{Gamma}(2, 1)$

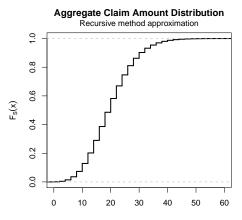
> Fs <- aggregateDist("recursive",</pre>

model.freq = "poisson", model.sev = fx, lambda = 10, x.scale = 2)

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Example (continued)

> plot(Fs)





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Example (continued)

> summary(Fs)

 Aggregate Claim Amount Empirical CDF:

 Min. 1st Qu.
 Median
 Mean 3rd Qu.
 Max.

 0.00000
 14.00000
 20.00000
 19.99996
 26.00000
 74.00000

> knots(Fs)

[1] 0 2 4 6 8 10 12 14 16 18 20 22 24 26 28 30 32 [18] 34 36 38 40 42 44 46 48 50 52 54 56 58 60 62 64 66 [35] 68 70 72 74

> Fs(c(10, 15, 20, 70))

[1] 0.1287553 0.2896586 0.5817149 0.9999979

Example (continued)

> mean(Fs)
[1] 19.99996
> VaR(Fs)
90% 95% 99%
30 34 42
> TVaR(Fs)
90% 95% 99%

35.99043 39.56933 46.97385

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One more thing...

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What If Recursions Do Not Start?

For example, in the Compound Poisson case

$$f_{\mathcal{S}}(\mathbf{0}) = e^{-\lambda(1-f_{\mathcal{C}}(\mathbf{0}))}$$

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- If λ is large, $f_{S}(0) = 0$ numerically
- One solution:
 - 1 divide λ by 2^n
 - 2 convolve resulting distribution *n* times with itself

Example

> summary(Fsc)

Aggregate Claim Amount Empirical CDF: Min. 1st Qu. Median Mean 3rd Qu. 0.00000 14.00000 20.00000 19.99997 26.00000 Max. 108.00000

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Concluding Remarks

- See cran.r-project.org/package=actuar for the package
- Package vignettes provide complete documentation
- Please cite the software in publications:

```
> citation(package = "actuar")
```

To cite actuar in publications use:

C. Dutang, V. Goulet and M. Pigeon (2008). actuar: An R Package for Actuarial Science. Journal of Statistical Software, vol. 25, no. 7, 1-37. URL http://www.jstatsoft.org/v25/i07 [...]

Contribute!