

Granular Loss Modeling With Copulae

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About the speaker



- **Michal Pešta**
- Assistant Professor
- Research areas: errors-in-variables, change point, bootstrap, loss reserving, oncology



- **Charles University, Prague**
- 670 years of history (founded in 1348)
- Faculty of Mathematics and Physics
- Department of Probability and Mathematical Statistics

Agenda

Aims

Methodology

Results

Conclusions



Agenda



Aims

Loss Reserving

Lifetime of a Claim

Two Main Goals

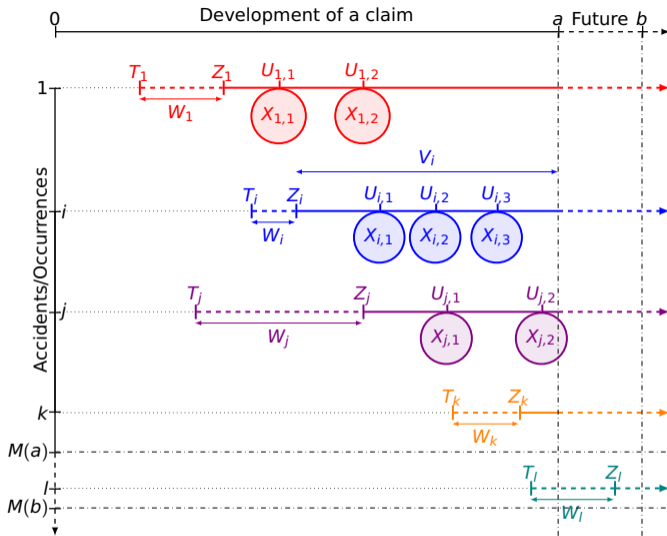


- **Primary**: predict future claim cash flows in non-life insurance and their uncertainty
- **Secondary**: back-predict incurred but not reported claims due to truncated data

Aggregated vs Granular

- Pitfalls of the conventional reserving techniques:
 - **loss of information** from the policy and the claim's development due to the aggregation, cf. Norberg (1993)
 - usually small number of observations in the triangle
 - only few observations for recent accident years
 - sensitivity to the most recent paid claims
- How to possibly overcome the issues:
 - individual/claim-by-claim/micro-level/**granular** data, which do not represent a mainstream in the reserving field, e.g., Antonio & Plat (2014)

Illustration



Agenda



Methodology

- Reporting Dates and Delays
- Payment Dates and Amounts
- Truncation of Accident Dates
- Dependence Between LoBs

- The time ordered reporting dates $\{Z_i\}_{i \in \mathbb{N}}$ are arrival times of a **non-homogeneous Poisson process** $\{M(t)\}_{t \geq 0}$ with a parametric intensity $\psi(t; \boldsymbol{\rho})$ such that

$$M(t) = \sum_{i=1}^{\infty} \mathbb{1}\{Z_i \leq t\}$$

- The cumulative intensity $\Psi(t; \boldsymbol{\rho}) := \int_0^t \psi(v; \boldsymbol{\rho}) dv$ diverges if $t \rightarrow \infty$

Reporting Delays

- The reporting delays W_i 's are independent random variables
- Sequence $\{W_i\}_{i \in \mathbb{N}}$ is stochastically independent of $\{Z_i\}_{i \in \mathbb{N}}$
- Given $Z_n = z$, W_n has a parametric density $f_W(\cdot, z; \boldsymbol{\theta})$

Payment Dates

- The time ordered payment delays $\{U_{i,1} - Z_i, U_{i,2} - Z_i, \dots\}$ of the i th claim are arrival times of a non-homogeneous Poisson process $\{N_i(t)\}_{t \geq 0}$
- $N_i(t) | W_i, Z_i$
- Processes $\{N_i(t)\}_{t \geq 0}$, $i = 1, 2, \dots$ are independent with a parametric intensity $\lambda_i(t; \boldsymbol{\nu}, \boldsymbol{\beta})$ such that

$$N_i(t) = \sum_{k=1}^{\infty} \mathbb{1}\{U_{i,k} - Z_i \leq t\}$$

- The cumulative intensity $\Lambda_i(t; \boldsymbol{\nu}, \boldsymbol{\beta}) := \int_0^t \lambda_i(v; \boldsymbol{\nu}, \boldsymbol{\beta}) dv$ converges if $t \rightarrow \infty$

Payment Amounts

- Sets of the payment amounts $\{X_{ij}\}_j$'s are independent random sequences
- Sequence $\{X_{ij}\}_j$ forms an **AR process**
- Sequence $\{\{X_{ij}\}_j\}_{i \in \mathbb{N}}$ is stochastically independent of $\{Z_i\}_{i \in \mathbb{N}}$
- Given $Z_n = z$, the first payment X_{n1} has a parametric density $f_X(\cdot, z; \boldsymbol{\zeta})$

Accident Dates

- Accident dates as **displacement** of the reporting dates
- Reporting dates are fully observed, accident dates are **truncated**
- The displacement theorem (Kingman, 1993) provides that accident dates T_i 's are arrival times of another non-homogeneous Poisson process with a parametric intensity

$$\mu(t; \boldsymbol{\rho}, \boldsymbol{\theta}) = \int_{\mathbb{R}} \psi(z; \boldsymbol{\rho}) f_W(t, z; \boldsymbol{\theta}) dz$$

- **Back-fit** the incurred but not reported claims

- Suppose that $Y_t^{(\ell)}$ is a loss amount for a time period t (e.g., month) and for a LoB ℓ
- Assume that the **dependence** between lines of business (LoBs) is modeled via a parametric copula (possibly time-varying in order to capture dynamic behavior)
- E.g., $\ell \in \{1, 2\}$... two LoBs (material damage and bodily injury)

$$\mathbb{P} \left[Y_t^{(1)} \leq y^{(1)}, Y_t^{(2)} \leq y^{(2)} \right] = \mathbf{C} \left(\mathbb{P} \left[Y_t^{(1)} \leq y^{(1)} \right], \mathbb{P} \left[Y_t^{(2)} \leq y^{(2)} \right]; \boldsymbol{\alpha}(t) \right)$$

By-product for Stochastic Theory



- Probabilistic framework for the **n.i.n.i.d.** observations
- **Time-varying** models in an unbalanced panel data setup
- **Maximum likelihood** estimators derived
- Proved **consistency** and **asymptotic normality** of the estimators
- **Justification** for usage of the method

Agenda

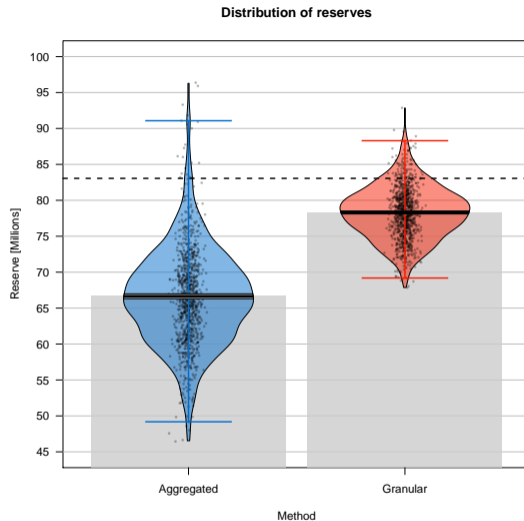


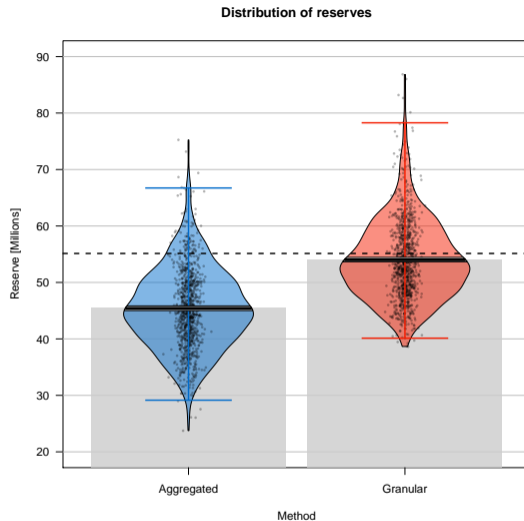
Results

Loss Reserves

Incurred But Not Reported

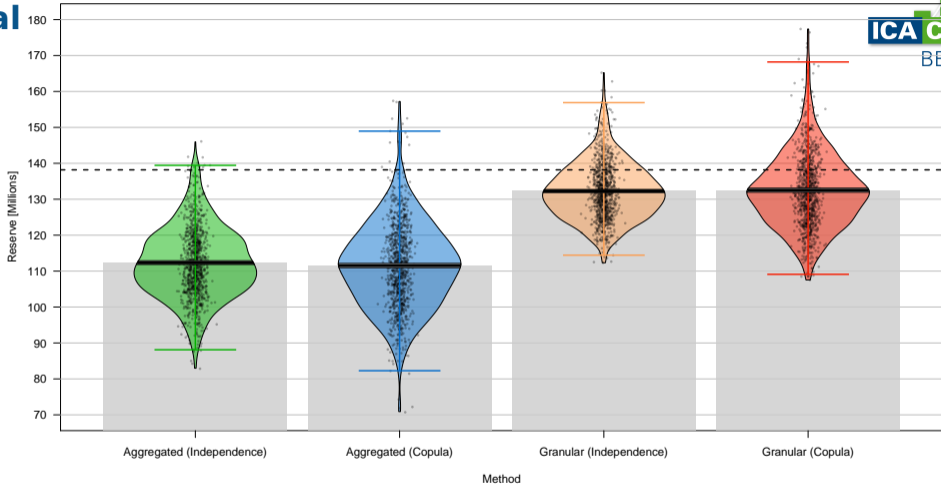
Material





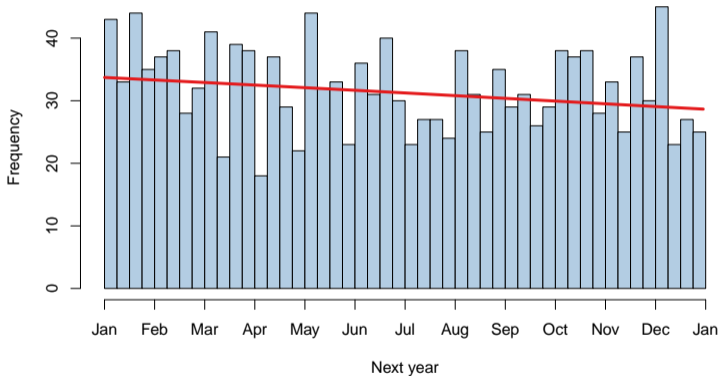
Total

Distribution of reserves



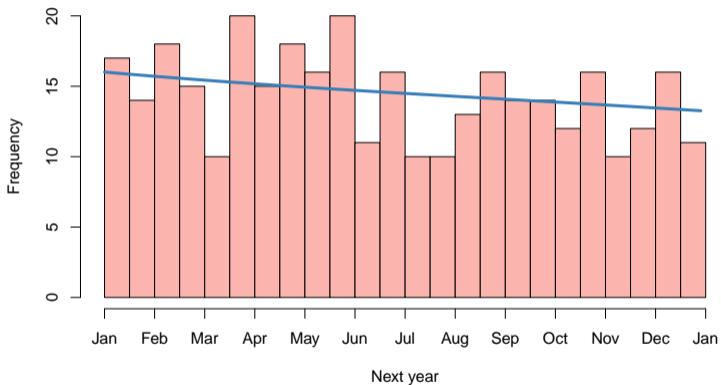
Predicted Future Reportings

Counts of reporting dates [Material]



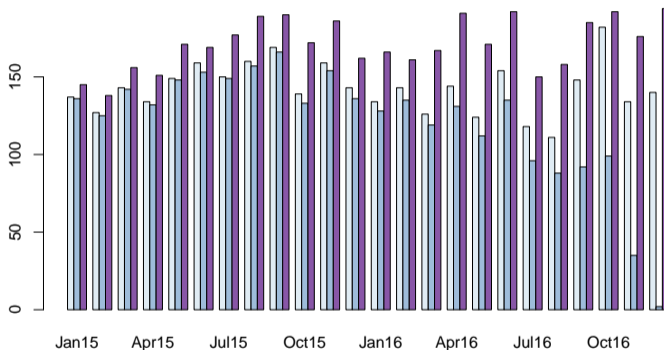
Predicted Future Reportings

Counts of reporting dates [Bodily]



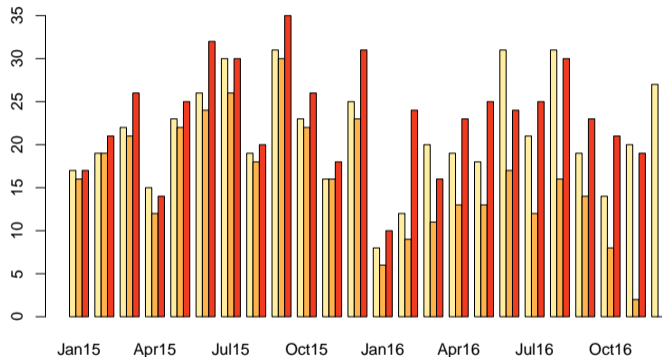
Back-fitted Recent Accidents

Counts of accident dates [Material]



Back-fitted Recent Accidents

Counts of accident dates [Bodily]



Agenda



Conclusions

References

Summary



- Focus on three synergic research areas
- Inventing stochastic methods for loss reserving based on claim-by-claim data
- Using a dynamic copula framework for modeling dependencies among types of claims
- Deriving appropriate statistical inference for these approaches

Thank you very much for your attention!

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References

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