



Application of the credibility principle in reinsurance pricing

David Raich
Angela Wünsche

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Agenda

1. Introduction into credibility theory
2. Some maths
3. Credibility for reinsurance pricing
4. Application – method used for MTPL
5. Vision

Introduction

Introduction



Initial situation:

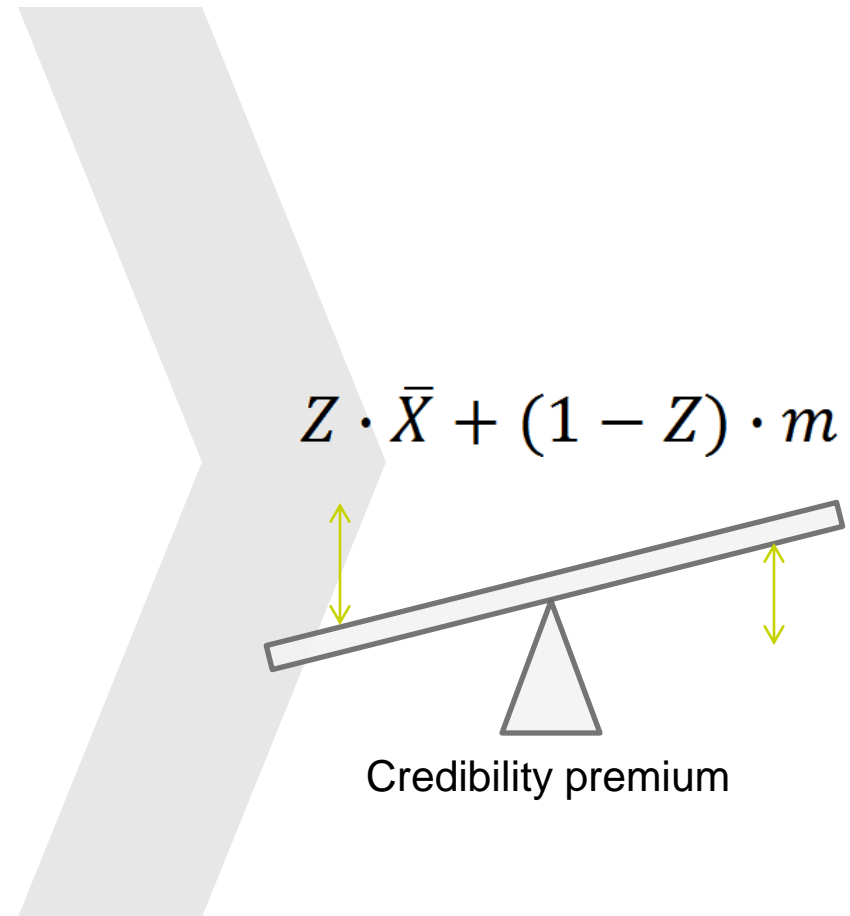
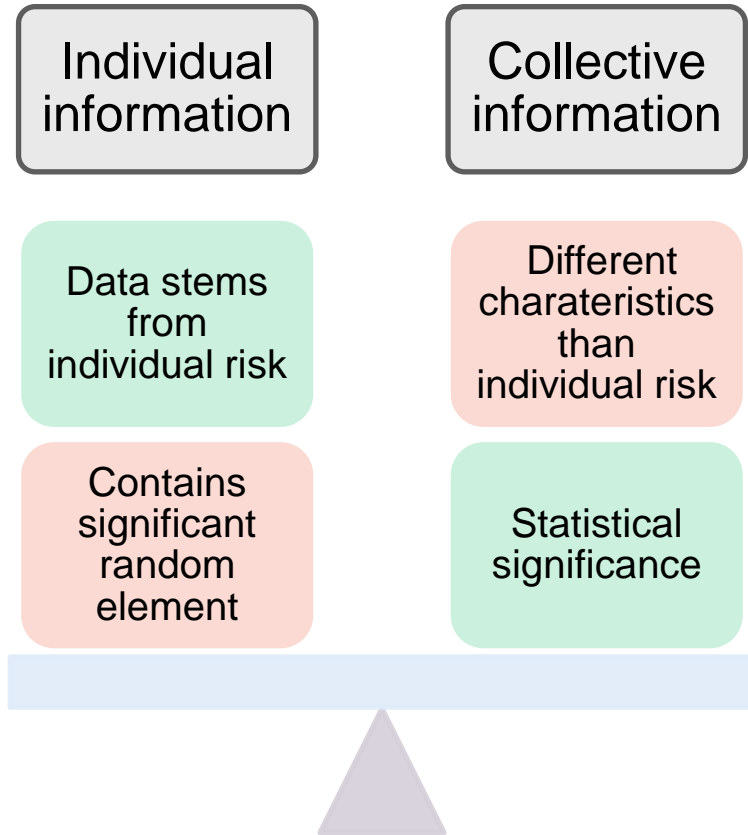
- Comprehensive information available for the collective (e.g. solid loss history or more)
- Limited data history available for individual risk

GOAL:

Make use of all (relevant) available information in order to get best estimate for the individual premium

Introduction

Collective vs individual information



Introduction

History of credibility theory

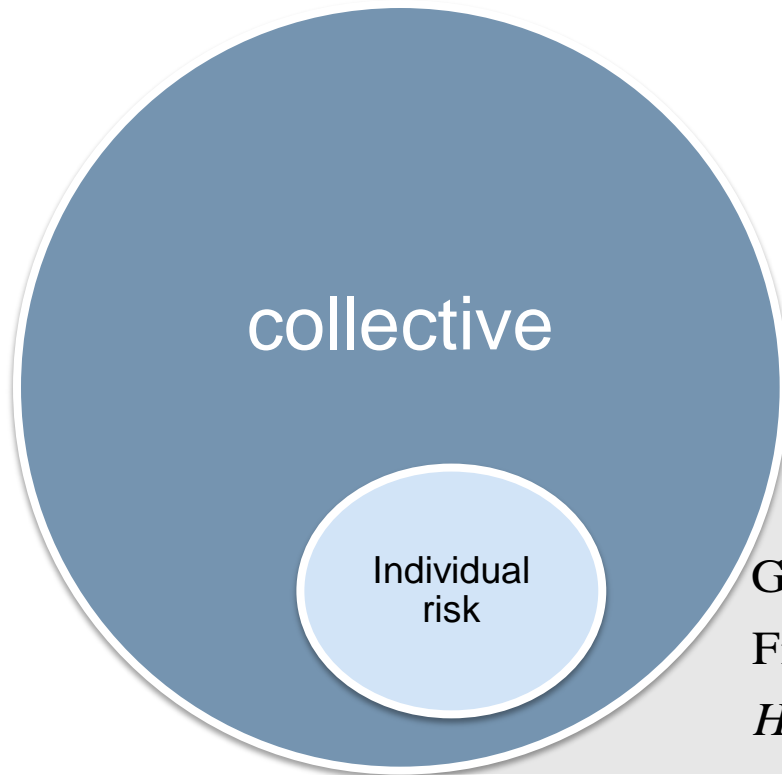
- Limited Fluctuation CT
 - Based on central limit theorem
 - Originally a “full or zero-credibility” method
 - Parameters in partial model introduced later calibrated according to actuarial judgement
- Greatest Accuracy CT
 - Heavily based on Bayesian statistics
 - “Best premium to charge”-approach
 - Results in stable and responsive estimator

Stability-oriented approach

Precision-oriented approach

Some maths

Mathematical Formulation



$$\mathbb{F} = \{F_{\mathcal{G}} \mid \mathcal{G} \in \Theta\}$$

Family of distributions indexed by risk profile \mathcal{G}

$$H[\Theta] = E[X \mid \Theta]$$

Individual premium

$$\mu = E[X] = E_{\Theta}(E[X \mid \Theta])$$

Collective premium

GOAL:

Given observations x_1, \dots, x_n for an individual risk $F_{\mathcal{G}}$

Find a good estimator for the individual premium

$$H(\mathcal{G}) = E[X_{n+1} / \mathcal{G}]$$

Main results – Bayesian estimator

$$\pi(\mathcal{G})$$

A priori density function of risk profile

$$f(x_1, \dots, x_n | \mathcal{G})$$

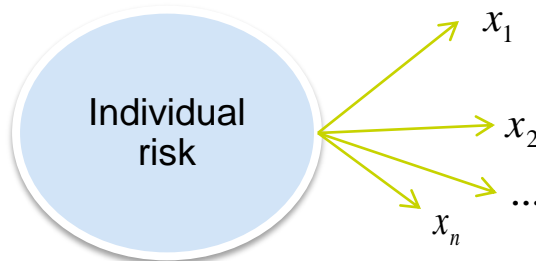
Conditional density function of losses

$$f(x_1, \dots, x_n, \mathcal{G})$$

Joint density function

$$H(\mathcal{G}) = \int x \cdot f(x | \mathcal{G}) dx$$

Individual Premium



$$\pi(\mathcal{G} | \mathbf{x})$$

A posteriori pdf of risk profile

$$P(B) = \sum_i P(B | A_i) \cdot P(A_i)$$

$$= \mathbf{x} \int H(\mathcal{G}) \pi(\mathcal{G} | \mathbf{x}) d\mathcal{G}$$

$$P(A, B) = P(A | B) \cdot P(B)$$



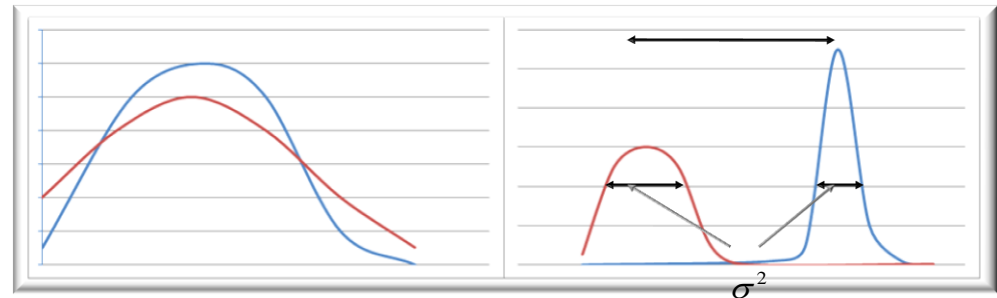
Main results – Bühlmann model

collective observations

$$\hat{\mu}(\Theta) = (1 - Z_n) \mu + Z_n \bar{X}_n$$

$$Z_n = \frac{n}{n + k}$$

$$k = \frac{\sigma^2}{a}$$



more weight assigned
to collective information

more weight assigned
to individual information

Main results – Bühlmann-Straub model

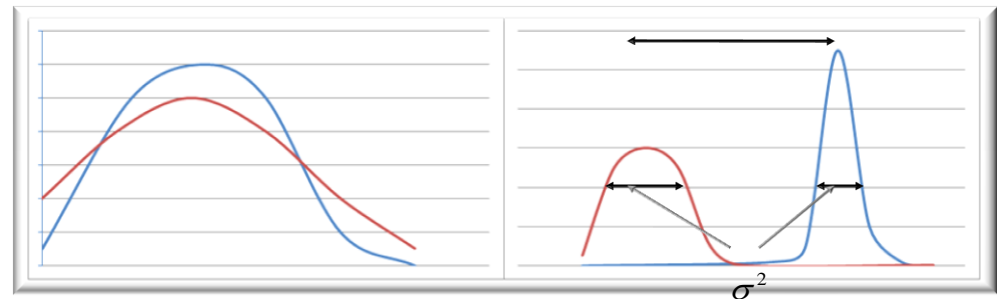
collective observations

$$\hat{\mu}^w(\Theta) = (1 - Z_n^w) \mu + Z_n^w \bar{X}_n^w$$

a

$$Z_n^w = \frac{w}{w + k}$$

$$k = \frac{\sigma^2}{a}$$



more weight assigned
to collective information

more weight assigned
to individual information

Credibility for reinsurance pricing

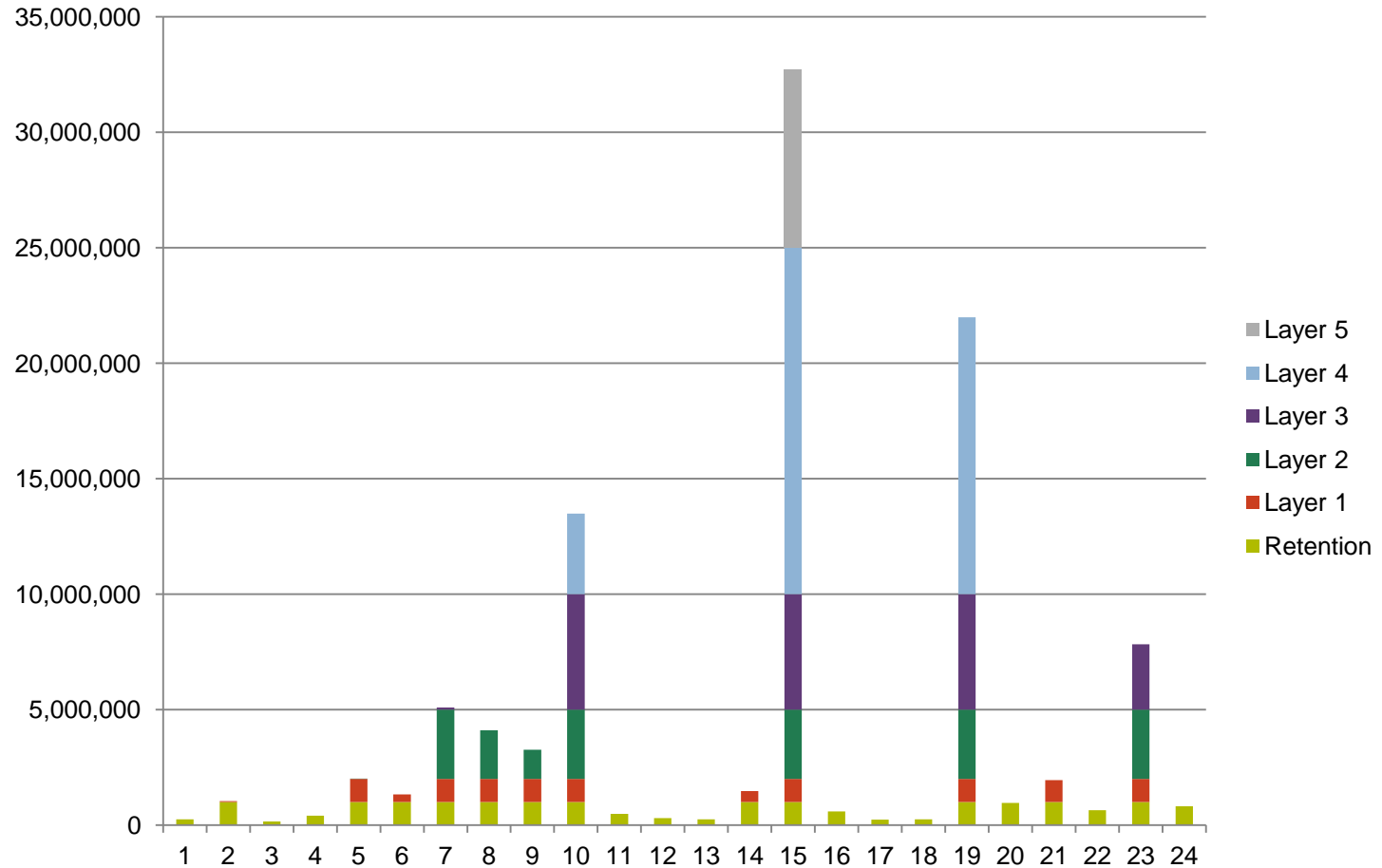
General comments

What problems do we face in reinsurance pricing?

- Pricing XL business for motor
- Usually data are only given back for the last 10 years
- Need to project losses to ultimate, where development can take much longer than 10 years
- Data are only available excess a threshold
- Hence scarce data, which may be insufficient to price a client based on experience
- We want to make use of all available data in market and weight a client against the market

naturally a application field of credibility

General comments



Challenges

Challenges:

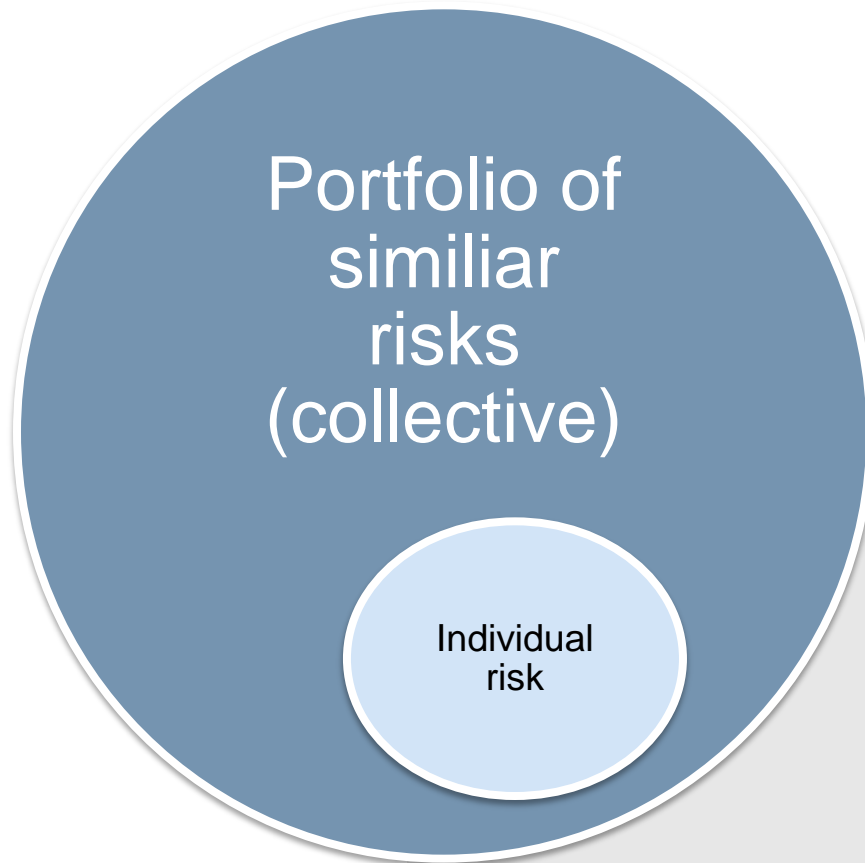
- Choice of appropriate portfolio
- Pure niche portfolios still require individual treatment
- Pricing of a layer 3 m xs 2 m is different than pricing ill xs 25 m
- Credibility weight needs to be calculated in dependency of claims size
- Credibility applied to frequency / severity or rate?
- Parameter uncertainty?
- Which is the appropriate exposure measure?

Requirements for credibility approach

- Produce reasonable results i.e. increase precision
- Ensure stability and responsiveness
- One model for all layers
- Easy to explain
- Ensure consistent approach within one market
- Application still allows for underwriting judgement

“Any credibility procedure requires the actuary to exercise informed judgment, using relevant information. The use of credibility procedures is not always a precise mathematical process” (Actuarial Standards board)

Credibility for reinsurance XoL pricing



Initial situation:

- Comprehensive information available for the collective (e.g. solid loss history or more)
- Limited data history available for individual risk

GOAL:

Make use of all (relevant) available information in order to get best estimate for the individual premium

Credibility for reinsurance XoL pricing



Initial situation:

- Net Market rate available (=Collective information)
- Limited loss history available for individual portfolio

GOAL:

Make use of both information in order to get good estimate for the individual net rate

Application – Method used for MTPL


Theoretical framework – Compound Gamma-Poisson

$$f_N(n|\theta) = \frac{\theta^n e^{-\theta}}{n!}$$

$$f_{\Theta}(\theta) = \frac{b^a \theta^{a-1} e^{-b\theta}}{\Gamma(a)}$$

The unconditional distribution of N is negative binomial with parameter $(a, b/(1+b))$.

$$\Rightarrow \hat{\mu}(\Theta_l) = (1 - Z_l) \bar{N}^{w,Z} + Z_l \bar{N}_l^w$$

$$Z_l = \frac{w_l}{w_l + b}$$


Estimation of the parameters- Compound Gamma model

Estimation of parameter b:

$$E[\Theta] = a/b = \frac{1}{m} \sum_{l=1}^m \lambda_l$$

$$\text{Var}[\Theta] = a/b^2 = \frac{1}{m-1} \sum_{l=1}^m (\lambda_l - \bar{\lambda})^2$$

$$\Rightarrow b = \frac{\bar{\lambda}}{\frac{1}{m-1} \sum_{l=1}^m (\lambda_l - \bar{\lambda})^2}$$

Process needs to
be repeated for
different
thresholds

Estimation of the parameters- Compound Gamma model

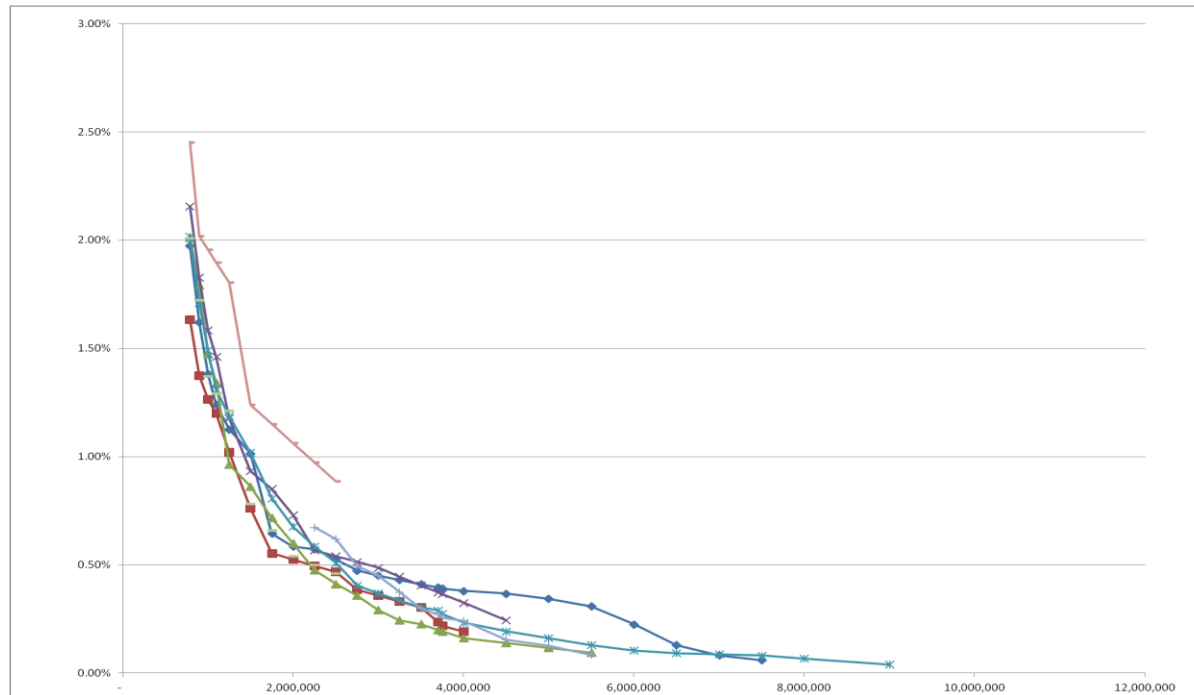
Estimation of parameter $b(T)$:

$$E[\Theta_T] = a(T) / b(T) = \frac{1}{m} \sum_{l=1}^m \lambda_l(T)$$

$$\text{Var}[\Theta_T] = a(T) / b(T)^2 = \frac{1}{m-1} \sum_{l=1}^m (\lambda_l(T) - \bar{\lambda}(T))^2$$

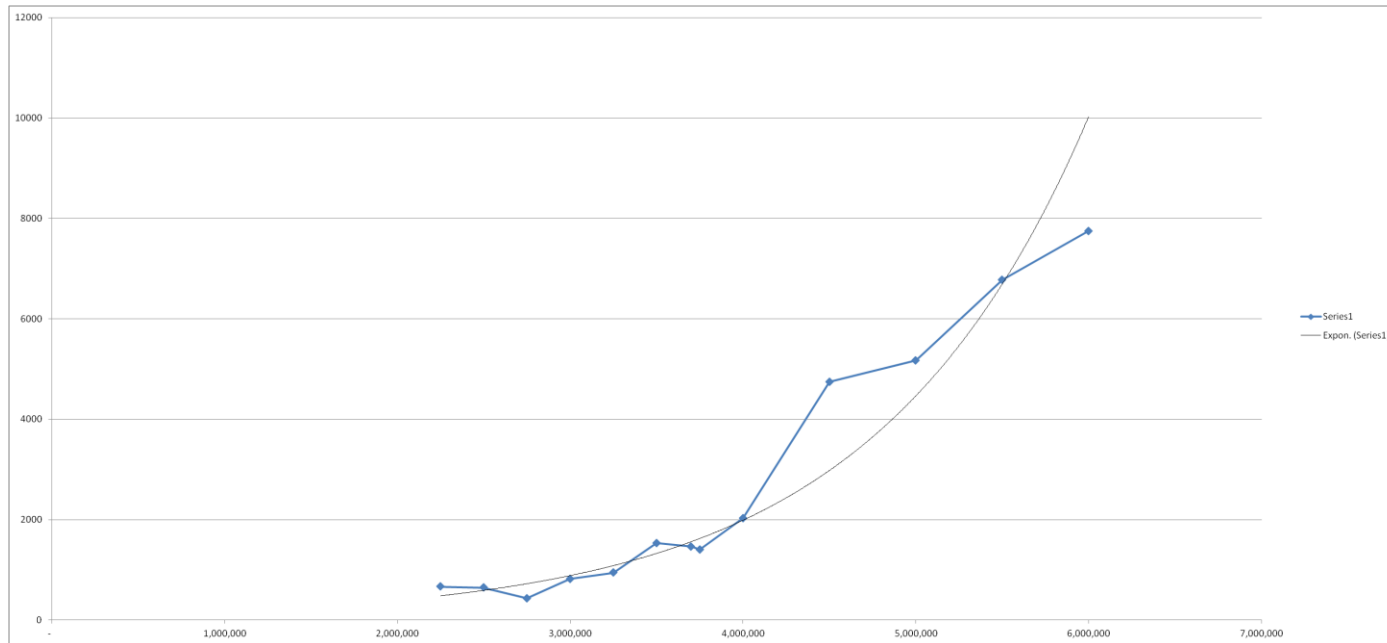
$$\Rightarrow b(T) = \frac{\bar{\lambda}(T)}{\frac{1}{m-1} \sum_{l=1}^m (\lambda_l(T) - \bar{\lambda}(T))^2}$$

Estimation of the parameters- Frequency of cedents



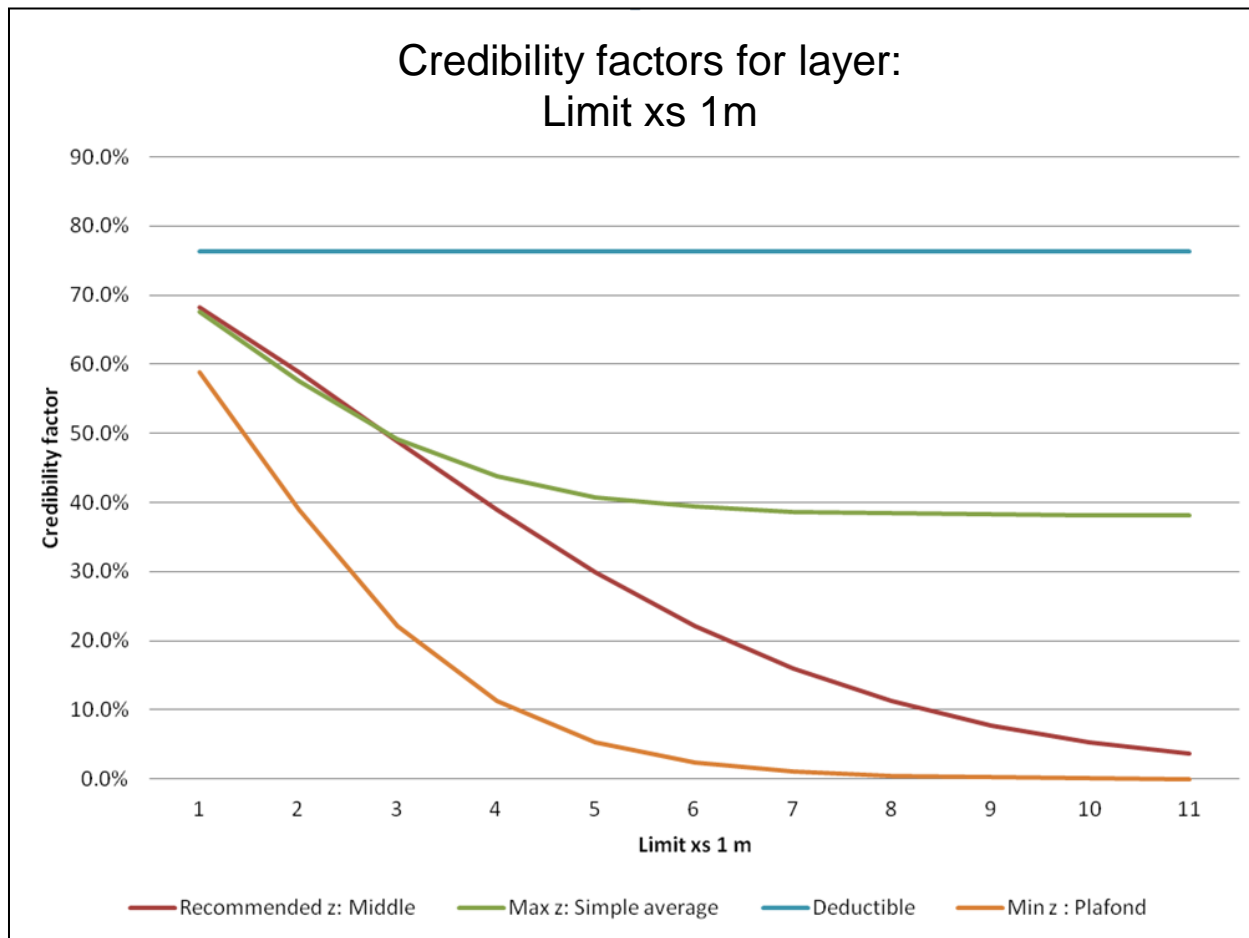
Expected Frequency of cedents @ different thresholds

Estimation of the parameters- Fit $b(T)$



Estimation of b for different thresholds incl. exponential fit

Application on client example



Status quo

- Where are we?
 - Credibility weight is calculated dependent on claim size and exposure
 - Calibrated on frequencies
 - Applied to the rate
 - Underwriting judgement is possible, because of the range given for the weight
 - Uncertainty of rate not explicitly taken into account, but within underwriting judgement

Vision

Vision

- Where to go?
 - Application for severity
 - Incorporation of market rate uncertainty
 - Expand application towards loadings (capital intensities)
- Other approaches in actuarial literature:
 - application on loss development factors (Pinot/Gogol)
 - making use of lower layers for upper layers

References

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Thank you for your attention