

Non-life insurance mathematics

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Overview

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Important issues	Models treated	Curriculum	Duration (in lectures)
What is driving the result of a non-life insurance company?	insurance economics models	Lecture notes	0,5
How is claim frequency modelled?	Poisson, Compound Poisson and Poisson regression	Section 8.2-4 EB	1,5
How can claims reserving be modelled?	Chain ladder, Bernhuetter Ferguson, Cape Cod,	Note by Patrick Dahl	2
How can claim size be modelled?	Gamma distribution, log-normal distribution	Chapter 9 EB	2
How are insurance policies priced?	Generalized Linear models, estimation, testing and modelling. CRM models.	Chapter 10 EB	2
Credibility theory	Buhlmann Straub	Chapter 10 EB	1
Reinsurance		Chapter 10 EB	1
Solvency		Chapter 10 EB	1
Repetition			1

Credibility theory was used before computers arrived

Motivation

Linear credibility

Example

Optimal credibility

The need for credibility theory

- So far we have studied how GLMs may be used to estimate relativities for rating factors with a moderate number of levels.
 - The rating factors were either categorical with few levels (e.g., gender), or
 - formed by grouping of a continuous variable (e.g., policy holder age)
- In case of insufficient data levels can be merged
- However, for categorical rating factors with a large number of levels without an inherent ordering there is no simple way to form groups to solve the problem with insufficient data

Example 1: car insurance

- Car model is an important rating factor in motor insurance
 - Exclusive cars are more attractive to thieves and more expensive to repair than common cars
 - Sports cars may be driven differently (more recklessly?) than family cars
- In Norway there are more than 3000 car model codes
- How should these be grouped?

Example 2: Experience rating

- The customer is used as a rating factor
- The *credibility estimators* are used to calibrate the pure premium
- The credibility estimators are weighted averages of
 - the pure premium based on the individual claims experience
 - The pure premium for the entire portfolio of insurance policies

Example 3: Geographic zones

- How should the premium be affected by the area of residence? (parallel to the car model example)
- Example: postcode. There are over 8000 postcodes in Norway
- We can merge neighbouring areas
- But they may have different risk characteristics

The credibility approach

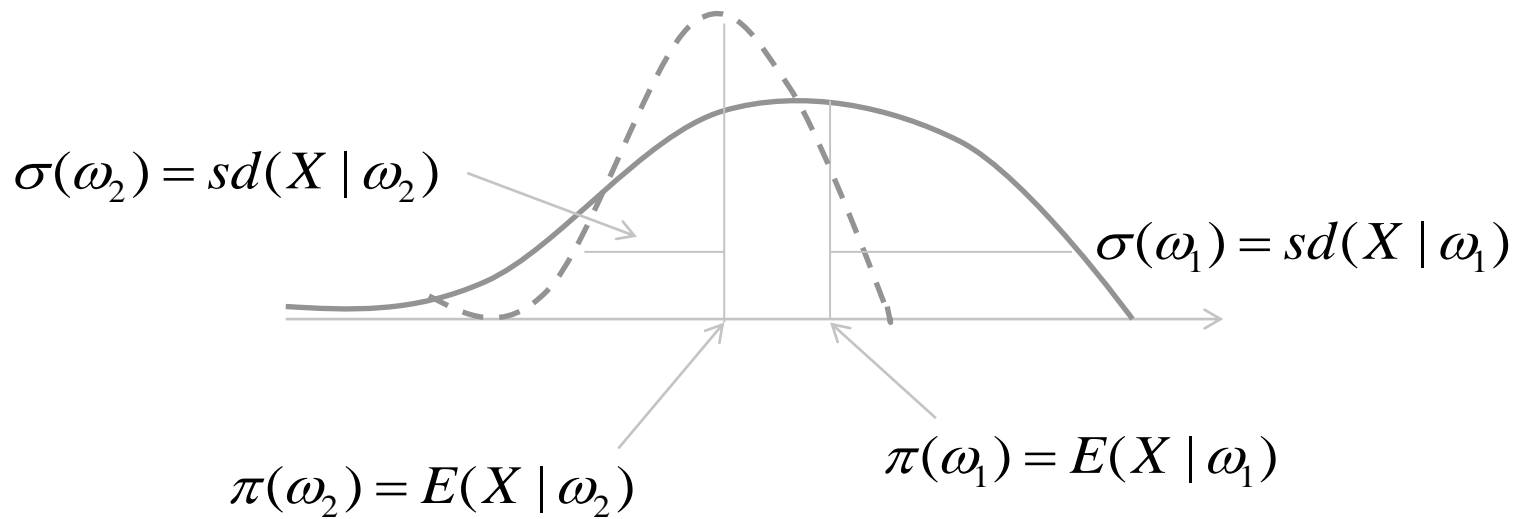
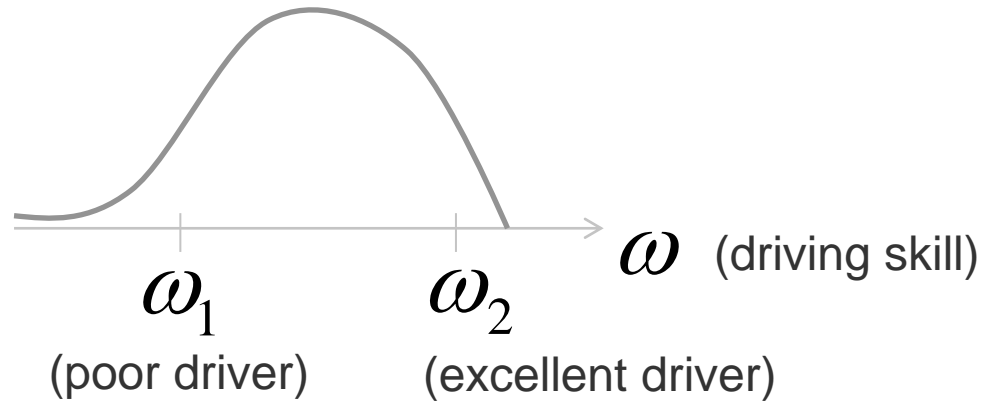
- The basic assumption is that policy holders carry a list of attributes with impact on risk
- The parameter ω could be how the car is used by the customer (degree of recklessness) or for example driving skill
- It is assumed that ω exists and has been drawn randomly for each individual
- X is the sum of claims during a certain period of time (say a year) and introduce

$$\pi(\omega) = E(X | \omega) \quad \text{and} \quad \sigma(\omega) = \text{sd}(X | \omega)$$

- We seek $\pi = \pi(\omega)$ the conditional pure premium of the policy holder as basis for pricing
- On group level there is a common ω that applies to all risks jointly
- We will focus on the individual level here, as the difference between individual and group is minor from a mathematical point of view

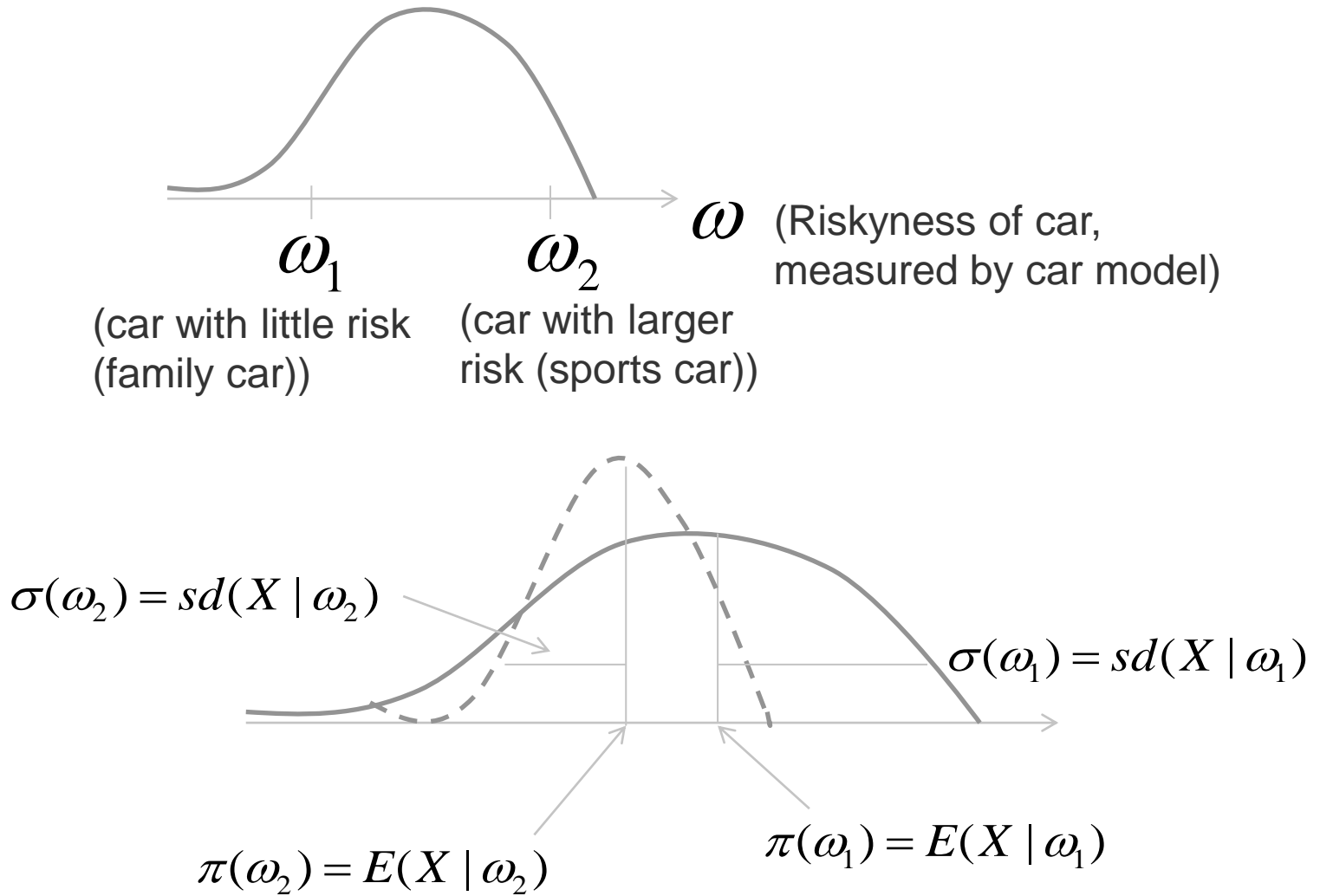
Omega is random and has been drawn for each policy holder

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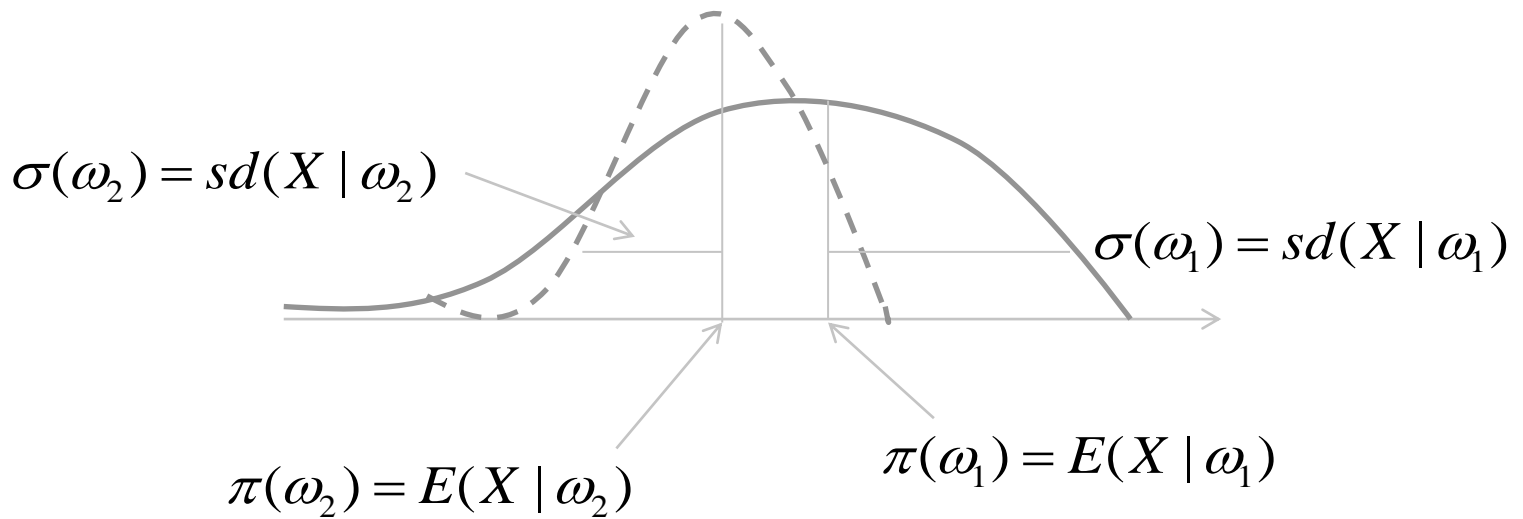
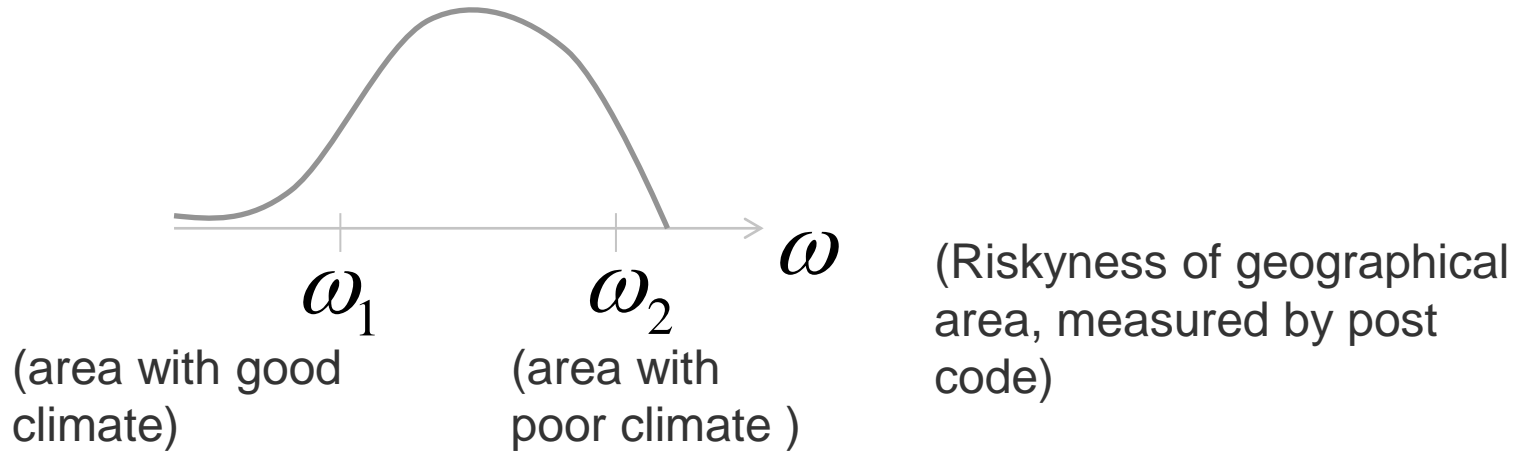
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Goal

We want to estimate

$$\bar{\pi}(\omega) = E\{E(X | \omega)\}, E\{\sigma^2(\omega)\} = E\{sd^2(X | \omega)\} \text{ and } \text{var}\pi(\omega)$$

These are called structural parameters.

If X_1, \dots, X_K are realizations from X , where X is the sum of claims during a year we want the following expression

$$\hat{\pi}_K = (1-w)\bar{\pi} + w\bar{X}_K$$

- We want to find what w is in two situations.
- The weight w defines a compromise between the average pure premium $\bar{\pi}$ of the population and the track record of the policy holder.

Linear credibility

- The standard method in credibility is the linear one with estimates of π of the form

$$\hat{\pi}_K = b_0 + b_1 X_1 + \dots + b_K X_K$$

where b_0, \dots, b_K are coefficients tailored to minimize the mean squared error $E(\hat{\pi}_K - \pi)^2$

- The fact that X, X_1, \dots, X_K are conditionally independent with the same distribution forces $b_1 = \dots = b_K$, and if w/K is their common value, the estimate becomes

$$\hat{\pi}_K = b_0 + w \bar{X}_K \quad \text{where} \quad \bar{X}_K = (X_1 + \dots + X_K) / K$$

Linear credibility

- To proceed we need the so-called *structural parameters*

$$\bar{\pi} = E\{\pi(\omega)\}, \quad \nu^2 = \text{var}\{\pi(\omega)\}, \quad \tau^2 = E\{\sigma^2(\omega)\}$$

where $\bar{\pi}$ is the average pure premium for the entire population.

- It is also the expectation for individuals since by the rule of double expectation

$$E(X) = E\{E\{X \mid \omega\}\} = E\{\pi(\omega)\} = \bar{\pi}$$

- Both ν and τ represent variation. The former is caused by diversity between individuals and the latter by the physical processes behind the incidents. Their impact on $\text{var}(X)$ can be understood through the rule of double variance, i.e.,

$$\text{var}(X) = E\{\text{var}(X \mid \omega)\} + \text{var}\{E(X \mid \omega)\}$$

$$= E\{\sigma^2(\omega)\} + \text{var}\{\pi(\omega)\} = \tau^2 + \nu^2$$

and ν^2 and τ^2 represent uncertainties of different origin that add to $\text{var}(X)$

Linear credibility

- The optimal linear credibility estimate now becomes

$$\hat{\pi}_K = (1-w)\bar{\pi} + w\bar{X}_K, \quad \text{where} \quad w = \frac{v^2}{v^2 + \tau^2 / K}$$

which is proved in Section 10.7, where it is also established that

$$E(\hat{\pi}_K - \pi) = 0 \quad \text{and that} \quad \text{sd}(\hat{\pi}_K - \pi) = \frac{v}{\sqrt{1 + Kv^2 / \tau^2}}.$$

- The estimate is unbiased and its standard deviation decreases with K
- The weight w defines a compromise between the average pure premium $\bar{\pi}$ of the population and the track record of the policy holder
- Note that $w=0$ if $K=0$; i.e. without historical information the best estimate is the population average

Estimation of structural parameters

- Credibility estimation is based on parameters that must be determined from historical data.
- Historical data for J policies that been in the company for K_1, \dots, K_J years are then of the form

$$\begin{array}{cccc|cc}
 1 & x_{11} & \dots & x_{1K_1} & \bar{x}_1 & s_1 \\
 \cdot & \cdot & & \cdot & \cdot & \cdot \\
 \cdot & \cdot & & \cdot & \cdot & \cdot \\
 \mathbf{J} & x_{J1} & \dots & x_{JK_1} & \bar{x}_J & s_J \\
 \text{Policies} & \text{Annual claims} & & & \text{mean} & \text{sd}
 \end{array}$$

where the j 'th row x_{j1}, \dots, x_{jK_j} are the annual claims from client j and \bar{x}_j and s_j their mean and standard deviation

- The following estimates are essentially due to Sundt (1983) and Bühlmann and Straub (1970):

Estimation of structural parameters

Let $K = K_1 + \dots + K_J$. Unbiased, moment estimates of the structural parameters are then

$$\hat{\pi} = \frac{1}{K} \sum_{j=1}^J K_j \bar{x}_j, \quad \hat{\tau}^2 = \frac{1}{K - J} \sum_{j=1}^J (K_j - 1) s_j^2$$

and

$$\hat{\nu}^2 = \frac{\sum_{j=1}^J (K_j / K) (\bar{x}_j - \hat{\pi})^2 - \hat{\tau}^2 (J - 1) / K}{1 - \sum_{j=1}^J (K_j / K)^2}$$

- For verification see Section 10.7.
- The expression for $\hat{\nu}^2$ may be negative. If it is let $\hat{\nu} = 0$ and assume that the underlying variation is too small to be detected.

The most accurate estimate

- Let X_1, \dots, X_K (policy level) be realizations of X dating K years back
- The most accurate estimate of π from such records is (section 6.4) the conditional mean

$$\hat{\pi}_K = E(X \mid x_1, \dots, x_K)$$

where x_1, \dots, x_K are the actual values.

- A natural framework is the common factor model of Section 6.3 where X, X_1, \dots, X_K are identically and independently distributed given ω
- This won't be true when underlying conditions change systematically
- A problem with the estimate above is that it requires a joint model for X, X_1, \dots, X_K and ω .
- A more natural framework is to break X down on claim number N and losses per incident Z .
- First linear credibility is considered

Optimal credibility

- The preceding estimates are the best *linear* methods but the Bayesian estimate $\hat{\pi}_K = E(X | x_1, \dots, x_K)$ is optimal among *all* methods and offers an improvement.
- Break the historical record x_1, \dots, x_K down on annual claim numbers n_1, \dots, n_K and losses z_1, \dots, z_n where $n = n_1 + \dots + n_K$ and assume independence.
- The Bayes estimate of $\pi = E(X) = E(N)E(Z)$ now becomes

$$\hat{\pi}_K = E(X | n_1, \dots, n_K, z_1, \dots, z_n) = E(N | n_1, \dots, n_K)E(Z | z_1, \dots, z_n)$$

and there are two parts.

- Often claim intensity μ fluctuates more strongly from one policy holder to another than does the expected loss $\xi_z = E(Z)$.
- We shall disregard all such variation in ξ_z .
- Then $\xi_z = E(Z | z_1, \dots, z_n)$ is independently of z_1, \dots, z_n and the Bayesian estimate becomes

$$\hat{\pi}_K = \xi_z E(N | n_1, \dots, n_K).$$

Optimal credibility

- A model for past and future claim numbers is needed. The natural one is the common factor model where N, N_1, \dots, N_k are conditionally independent and identically distributed given μ with N and all N_k being Poisson (μT).
- For μ assume the standard representation

$$\mu = \xi_\mu G \quad \text{and} \quad G \sim \text{Gamma}(\alpha)$$

where $E(G)=1$. It is shown in Section 10.7 that the estimate

$$\hat{\pi}_K = \xi_z E(N | n_1, \dots, n_K) \quad \text{now becomes}$$

$$\hat{\pi}_K = \bar{\pi} \frac{\bar{n} + \alpha / K}{\xi_\mu T + \alpha / K} \quad \text{where} \quad \bar{\pi} = \xi_\mu T \xi_z$$

$$\text{and } \bar{n} = (n_1 + \dots + n_K) / K$$

- Note that the population average $\bar{\pi}$ is adjusted up or down according to whether the average claim number \bar{n} is larger or smaller than its expectation $\xi_\mu T$. The error (proved in Section 10.7) is

$$E(\hat{\pi}_K - \pi) = 0 \quad \text{and} \quad \text{sd}(\hat{\pi}_K - \pi) = \frac{\bar{\pi}}{\sqrt{\alpha + K \xi_\mu T}}$$