

A Generic Claims Reserving Model

A Fundamental Risk Analysis

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Outline

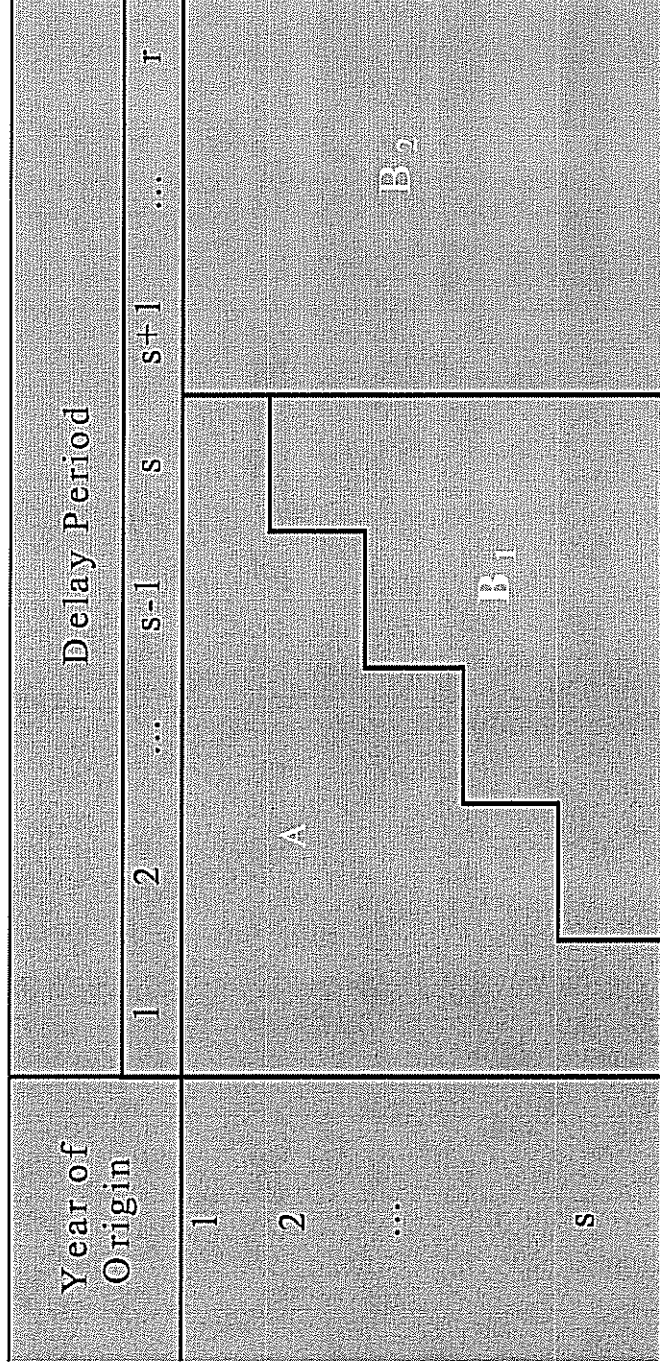
- Overview of the chain ladder's assumptions
- Generic claims reserving model
- Actuarial risk characteristics
- Advantages of the generic model
- More complex models
- Predicted mean square errors

The chain ladder

- Predictive Scope
- Assumptions (Kremer (1982))
- Conclusions

Chain ladder

Predictive Scope limited to B_1



Chain ladder - Assumptions

(Kremer (1982))

$$y_{w,j}^{***} = \ln y_{w,j}^* = \mu + \alpha_w + \beta_j + e_{w,j}$$

$$e_{w,j} \sim N(0, \sigma^2)$$

$$\sum_{w=1}^s \alpha_w = \sum_{j=1}^s \beta_j = 0$$

Chain ladder - Assumptions

(Alternative expressions for Kremer (1982))

$$\hat{C}_{w,s}^* \equiv \exp(\mu + \alpha_w) \left(\sum_{k=1}^s \exp(\beta_k) \right)$$

$$\text{If } \hat{P}_j^* \equiv \begin{cases} \exp(\beta_j) \left(\sum_{k=1}^s \exp(\beta_k) \right)^{-1} & j \leq s \\ 0 & j > s \end{cases}$$

$$\text{Then } E(y_{w,j}^*) = \hat{C}_{w,s}^* \hat{P}_j^*$$

The chain ladder - Conclusion

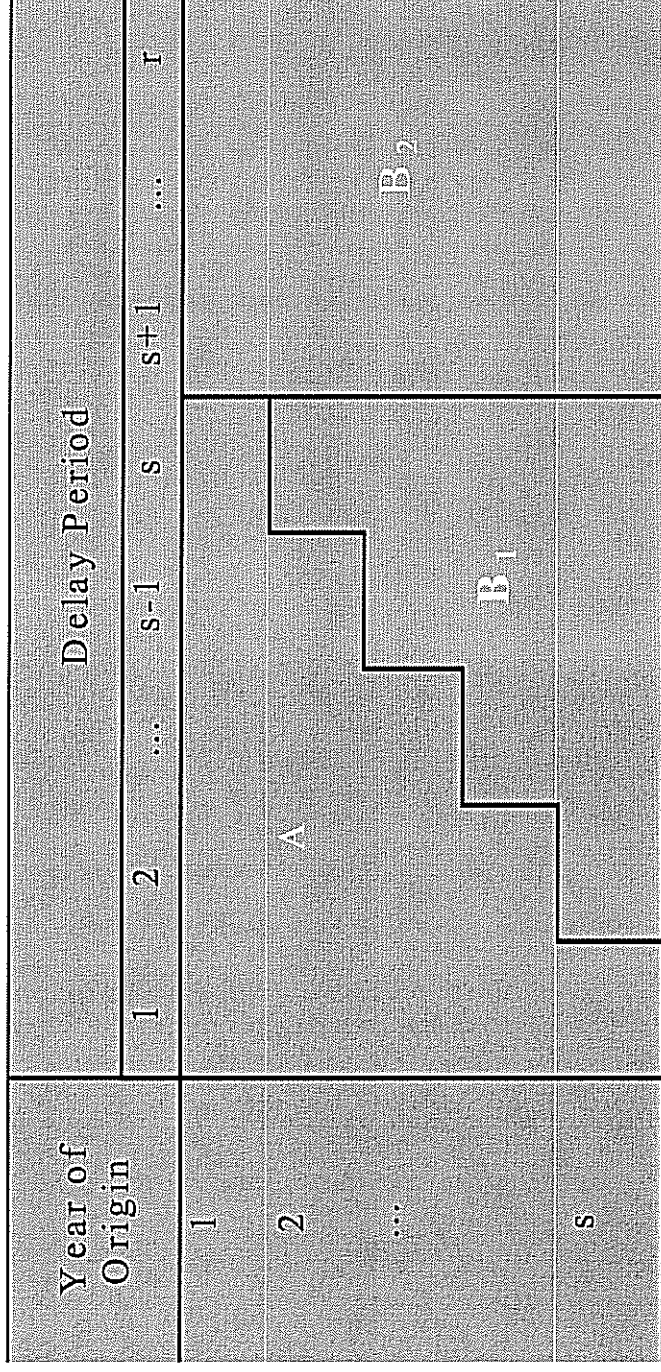
- Each term of the claims array can be expressed in terms of the marginal estimates.
- To address the limitations of the chain ladder and related methods
 - The row estimates $\hat{C}_{w,s}^*$ should be independent of s .
 - \hat{p}_j^* should be defined for all values of j

Generic claims reserving model

- Predictive Scope
- Assumptions
- Conclusions

Generic claims reserving model

Predictive Scope: B_1 and B_2



Generic model - Assumptions

- Probability density function $\pi(w, t)$
- Percentage cash flow and related functions
- Definition of $\pi(w, t)$
- Mean response and IBNR

Generic model - Assumptions

Probability density function $\pi(w, t)$.

$$\pi(w, t) \geq 0 \quad \forall t$$

$$\Pi(w, t) = \int_{z=0}^t \pi(w, z) dz = \Pr[z \leq t]$$

$$\int_{z=0}^{\infty} \pi(w, z) dz = 1$$

Generic model

Percentage cash flow and related functions

$$P(w, t_j) = \int_{z=0}^{t_j} \pi(w, z) dz$$

Cumulative
Percentage Cash Flow

$$S(w, t_j) = \int_{z=t_j}^{t_e} \pi(w, z) dz = 1 - P(w, t_j)$$

$$p(w, j) = \int_{z=t_{j-1}}^{t_j} \pi(w, z) dz$$

Incremental
Percentage Cash Flow

Generic model

Definition of $\pi(w, t)$

$$\pi(w, z) = g_w(\cdot)G(w, \hat{\lambda}_w, z)$$

$$\text{If } g_w^{-1}(\cdot) = \int_{z=0}^{\infty} G(w, \hat{\lambda}_w, z) dz$$

$$f(j, \hat{\lambda}_w) = \int_{z=j-1}^{z=j} G(w, \hat{\lambda}_w, z) dz$$

$$\text{Then } \begin{cases} C_w = \exp(\alpha_w) g_w^{-1}(\cdot) \\ p(w, j) = f(j, \hat{\lambda}_w) g_w(\cdot) \end{cases}$$

Generic model - Assumptions

Mean response and IBNR

Incremental
model:

$$E(y(w, j)) = \exp(\alpha_w) f(j, \hat{\lambda}_w)$$

Cumulative
model:

$$E(Y(w, t)) = \exp(\alpha_w) \int_{z=0}^t G(w, \hat{\lambda}_w, z) dz$$

IBNR:

$$IBNR_{(w, s-w+1)} = \exp(\alpha_w) \int_{z=t}^{\infty} G(w, \hat{\lambda}_w, z) dz$$

Actuarial risk characteristics

- Depending on the moments of $\pi(w, t)$
 - Skewness
 - Kurtosis
- Hazard rate

$$h(w, t) = \frac{\left(\frac{\partial \Pi(w, z)}{\partial z} \right)_{z=t}}{1 - \Pi(w, t)} = - \left(\frac{\partial \left(\log(1 - \Pi(w, z)) \right)}{\partial z} \right)_{z=t}$$

Advantages of the generic model

- Predictive scope
- Basis of innumerable reserving models
- Delay for a given IBNR and inverse function can be obtained

$$t = \Pi^{-1} \left(1 - \frac{IBNR_{(w,t)}}{C_w} \right) = \Pi^{-1} (P(w, t))$$

- Extracts characteristics of actuarial risk
- A platform for various important portfolio studies

More complex models

$$\text{If } E(Y(w, j)) = \sum_{k=1}^n C_{kw} P_k(w, j) = C_w P(w, j)$$

$$\left\{ \begin{array}{l} C_w = \sum_{k=1}^n C_{kw} \\ P(w, j) = \sum_{k=1}^n \vee_{w_k} P_k(w, j) \\ h(w, j) = \sum_{k=1}^n \wedge_{w_k} h_k(w, j) \end{array} \right\} \text{ then } \left\{ \begin{array}{l} \vee_{w_i} = \frac{C_{iw}}{\sum_{k=1}^n C_{kw}} \\ \wedge_{w_i} = \frac{IBNR_{(w,j)_i}}{\sum_{k=1}^n IBNR_{(w,j)_i}} \end{array} \right\} \text{ where}$$

Predicted mean square errors

Let $z(w, j)$ be the unknown future incremental claims in development year j . Mean square errors are derived for payment year totals and for

$$E(z(w, j)) = \exp(\hat{\lambda}_w) \int_{j-1}^j G(w, \beta_w, t) dt$$

$$E\left(\sum_{j=s-1}^{s-1} z(w, j)\right) = IBNR_{(w, s-1)}$$

$$E\left(\sum_{w=1}^u \sum_{j=s-1}^{s-1} z(w, j)\right) = IBNR_s$$

Predicted mean square errors

Comments

The derived mean square errors are intended for

- incremental claims reserving models
- models with simple variance/covariance structures that can be easily extracted from the model

The formulation of the mean response drastically simplifies the mean square error calculations

Bayesian Markov chain Monte Carlo methods offer a simpler alternative approach