

Non-Life Insurance: Mathematics and Statistics

Exercise sheet 12

Exercise 12.1 (Inhomogeneous) Credibility Estimators for Claim Counts

Suppose that in Table 1 we are given the current year's claim counts data for 5 different regions, where, for all $i \in \{1, \dots, 5\}$, $v_{i,1}$ denotes the number of policies in region i and $N_{i,1}$ the number of claims in region i . We assume that we are in the Bühlmann-Straub model framework with $I = 5$, $T = 1$ and

$$N_{i,1} | \Theta_i \sim \text{Poi}(\mu(\Theta_i)v_{i,1}),$$

with $\mu(\Theta_i) = \Theta_i \lambda_0$ and $\lambda_0 = 0.088$, for all $i \in \{1, \dots, 5\}$. Moreover, we assume that the pairs $(\Theta_1, N_{1,1}), \dots, (\Theta_5, N_{5,1})$ are independent and $\Theta_1, \dots, \Theta_5$ are i.i.d. with $\mathbb{E}[\Theta_1] = 1$ and $\Theta_1 > 0$ a.s. Finally, we set $\tau^2 = \text{Var}(\mu(\Theta_1)) = 0.00024$.

region i	$v_{i,1}$	$N_{i,1}$
1	50'061	3'880
2	10'135	794
3	121'310	8'941
4	35'045	3'448
5	4'192	314

Table 1: Observed numbers of policies $v_{i,1}$ and numbers of claims $N_{i,1}$ in the 5 regions.

- (a) Calculate the inhomogeneous credibility estimator $\widehat{\mu(\Theta_i)}$ for each region $i \in \{1, \dots, 5\}$ and comment on the results. What would we observe if we decreased the volatility τ^2 between the risk classes?
- (b) We denote next year's numbers of policies by $v_{1,2}, \dots, v_{5,2}$ and next year's numbers of claims by $N_{1,2}, \dots, N_{5,2}$. Suppose that $N_{i,1}$ and $N_{i,2}$ are independent, conditionally given Θ_i , for all $i \in \{1, \dots, 5\}$, and that the number of policies grows 5% in each region. For all $i \in \{1, \dots, 5\}$, under the assumption $N_{i,2} | \Theta_i \sim \text{Poi}(\mu(\Theta_i)v_{i,2})$, calculate the mean square error of prediction

$$\mathbb{E} \left[\left(\frac{N_{i,2}}{v_{i,2}} - \widehat{\mu(\Theta_i)} \right)^2 \right].$$

Exercise 12.2 (Homogeneous) Credibility Estimators for Claim Sizes

Suppose that in Table 2 we are given claim size data for two different years and four different risk classes, where $v_{i,t}$ denotes the number of claims in risk class i and year t and $Y_{i,t}$ the total claim size in risk class i and year t , for all $i \in \{1, 2, 3, 4\}$ and $t \in \{1, 2\}$. We assume that we are in the Bühlmann-Straub model framework with $I = 4$, $T = 2$ and

$$Y_{i,t} | \Theta_i \sim \Gamma(\mu(\Theta_i)cv_{i,t}, c),$$

with $\mu(\Theta_i) = \Theta_i$ and $c > 0$, for all $i \in \{1, 2, 3, 4\}$ and $t \in \{1, 2\}$. Moreover, we assume that $(\Theta_1, Y_{1,1}, Y_{1,2}), \dots, (\Theta_4, Y_{4,1}, Y_{4,2})$ are independent and that $\Theta_1, \dots, \Theta_4$ are i.i.d. with $\mathbb{E}[\Theta_1^2] < \infty$ and $\Theta_1 > 0$ a.s. Finally, we also assume that $Y_{i,1}$ and $Y_{i,2}$ are independent, conditionally given Θ_i , for all $i \in \{1, 2, 3, 4\}$.

risk class i	$v_{i,1}$	$Y_{i,1}$	$v_{i,2}$	$Y_{i,2}$
1	1'058	8'885'738	1'111	13'872'665
2	3'146	7'902'445	3'303	4'397'183
3	238	2'959'517	250	6'007'351
4	434	10'355'286	456	15'629'998

Table 2: Observed numbers of claims $v_{i,1}$ and $v_{i,2}$ and total claim sizes $Y_{i,1}$ and $Y_{i,2}$ in the 4 risk classes.

- (a) Calculate the homogeneous credibility estimator $\widehat{\mu(\Theta_i)}^{\text{hom}}$ for each risk class $i \in \{1, 2, 3, 4\}$ and comment on the results.
- (b) We denote next year's numbers of claims by $v_{1,3}, \dots, v_{4,3}$ and next year's total claim sizes by $Y_{1,3}, \dots, Y_{4,3}$. Suppose that $Y_{i,1}, Y_{i,2}$ and $Y_{i,3}$ are independent, conditionally given Θ_i , for all $i \in \{1, 2, 3, 4\}$, and that the number of claims grows 5% in each risk cell. For all $i \in \{1, 2, 3, 4\}$, under the assumption $Y_{i,3} | \Theta_i \sim \Gamma(\mu(\Theta_i) c v_{i,3}, c)$, estimate the mean square error of prediction

$$\mathbb{E} \left[\left(\frac{Y_{i,3}}{v_{i,3}} - \widehat{\mu(\Theta_i)}^{\text{hom}} \right)^2 \right].$$

Exercise 12.3 Degenerate MLE and the Poisson-Gamma Model

Suppose that in a given line of business we observed the following claim counts data for $T = 5$ years $t = 1, \dots, T$:

t	1	2	3	4	5
N_t	0	0	0	0	0
v_t	10	10	10	10	10

Table 3: Observed claim counts N_t and corresponding volumes v_t for $T = 5$ years $t = 1, \dots, T$.

- (a) First, we assume a Poisson model for the claim counts, i.e. N_1, \dots, N_T are independent with $N_t \sim \text{Poi}(\lambda v_t)$, $t = 1, \dots, T$, for an unknown claim frequency parameter $\lambda > 0$. Calculate the MLE $\widehat{\lambda}_T$ of λ . Does this estimate $\widehat{\lambda}_T$ make sense for premium calculation?
- (b) Now we assume a Poisson-gamma model for the claim counts, i.e. $\Lambda \sim \Gamma(\gamma, c)$ with $\gamma = 1$ and $c = 50$, and, conditionally given Λ , N_1, \dots, N_T are independent with $N_t \sim \text{Poi}(\Lambda v_t)$, $t = 1, \dots, T$.

- (i) Determine the prior estimator λ_0 and the posterior estimator $\widehat{\lambda}_T^{\text{post}}$, conditionally given data $(N_1, v_1), \dots, (N_T, v_T)$, of the unknown parameter Λ .
- (ii) Find the credibility weight $\alpha_T \in (0, 1)$ such that

$$\widehat{\lambda}_T^{\text{post}} = \alpha_T \widehat{\lambda}_T + (1 - \alpha_T) \lambda_0.$$

- (iii) Suppose we have an additional observation $(N_{T+1}, v_{T+1}) = (1, 10)$ within the Poisson-gamma model framework and that $\widehat{\lambda}_{T+1}^{\text{post}}$ denotes the posterior estimator, conditionally given data $(N_1, v_1), \dots, (N_{T+1}, v_{T+1})$, of the unknown parameter Λ . Find the credibility weight $\beta_{T+1} \in (0, 1)$ such that

$$\widehat{\lambda}_{T+1}^{\text{post}} = \beta_{T+1} \frac{N_{T+1}}{v_{T+1}} + (1 - \beta_{T+1}) \widehat{\lambda}_T^{\text{post}}.$$

- (c) Finally, we assume a Poisson-normal model for the claim counts, i.e. $\Lambda \sim \mathcal{N}(\mu, \sigma^2)$ with mean $\mu \in \mathbb{R}$ and variance $\sigma^2 > 0$, and, conditionally given Λ , N_1, \dots, N_T are independent with $N_t \sim \text{Poi}(\Lambda v_t)$, $t = 1, \dots, T$. Is such a model reasonable?

Exercise 12.4 Pareto-Gamma Model

Suppose that $\Lambda \sim \Gamma(\gamma, c)$ with prior shape parameter $\gamma > 0$ and prior scale parameter $c > 0$ and, conditionally given Λ , the components of $\mathbf{Y} = (Y_1, \dots, Y_T)$ are independent with $Y_t \sim \text{Pareto}(\theta, \Lambda)$ for some threshold $\theta > 0$, for all $t \in \{1, \dots, T\}$.

- (a) Show that the posterior distribution of Λ , conditional on \mathbf{Y} , is given by

$$\Lambda | \mathbf{Y} \sim \Gamma \left(\gamma + T, c + \sum_{t=1}^T \log \frac{Y_t}{\theta} \right).$$

- (b) For the estimation of the unknown tail index parameter Λ of the Pareto distributions, we define the prior estimator $\lambda_0 = \mathbb{E}[\Lambda]$ and the observation based estimator (MLE of the Pareto tail index parameter)

$$\hat{\lambda}_T = \frac{T}{\sum_{t=1}^T \log \frac{Y_t}{\theta}}.$$

Find the credibility weight $\alpha_T \in (0, 1)$ such that the posterior estimator $\hat{\lambda}_T^{\text{post}} = \mathbb{E}[\Lambda | \mathbf{Y}]$ has the credibility form

$$\hat{\lambda}_T^{\text{post}} = \alpha_T \hat{\lambda}_T + (1 - \alpha_T) \lambda_0.$$

- (c) Show that for the (conditional mean square error) uncertainty of the posterior estimator $\hat{\lambda}_T^{\text{post}}$ we have

$$\mathbb{E} \left[\left(\Lambda - \hat{\lambda}_T^{\text{post}} \right)^2 \middle| \mathbf{Y} \right] = (1 - \alpha_T) \frac{1}{c} \hat{\lambda}_T^{\text{post}}.$$

- (d) Let $\hat{\lambda}_{T-1}^{\text{post}}$ denote the posterior estimator in the sub-model where we only have observed (Y_1, \dots, Y_{T-1}) . Find the credibility weight $\beta_T \in (0, 1)$ such that the posterior estimator $\hat{\lambda}_T^{\text{post}}$ has the recursive update structure

$$\hat{\lambda}_T^{\text{post}} = \beta_T \frac{1}{\log \frac{Y_T}{\theta}} + (1 - \beta_T) \hat{\lambda}_{T-1}^{\text{post}}.$$