

Lecture (30)

Bayesian Premium & Bühlmann Premium estimator

Example 17.10 p. 413

* Revis lectures (28) & (29)

(EX. 17.8 continued) Determine the Bayesian premium
p. 410

Ans: ∴ the amount of a claim has an exponential distⁿ with mean $1/\theta$ (see EX 17.3 p. 403)

$$\therefore \mu(\theta) = \theta^{-1} = \frac{1}{\theta}$$

For Bayesian premium estimate, we can use Eq. (7) in lecture (29).

$$E(X_{n+1} | X=x) = \int \mu(\theta) \pi_{n+1}(\theta | X) d\theta$$

From EX 17.8

$$\pi(\theta | 100, 950, 450) = \frac{\theta^6 e^{-2500\theta} (2500)^7}{720}$$

$$\begin{aligned} \therefore E(X_4 | 100, 950, 450) &= \int \frac{1}{\theta} \frac{\theta^6 e^{-2500\theta} (2500)^7}{720} d\theta \\ &= \frac{2500^7}{720} \int_0^{\infty} \theta^5 e^{-2500\theta} d\theta \\ &= \frac{2500^7}{720} \int_0^{\infty} \left(\frac{u}{2500}\right)^5 e^{-\frac{u}{2500}} \frac{du}{2500}, \quad u = 2500\theta \\ &= \frac{2500}{720} \int_0^{\infty} u^5 e^{-u} du \\ &= \frac{2500}{720} \Gamma(6) = \frac{2500(5!)}{720} \end{aligned}$$

$$\therefore E(X_4 | 100, 950, 450) = \frac{2500(120)}{720} = 416.67$$

This result could also have been obtained as follows.

From EX 17.8, the predictive density is

$$f(x_4 | 100, 950, 450) = \frac{7(2500)^7}{(2500 + x_4)^8} \quad \text{which is a pareto density with parameters } (\alpha, \theta) \Rightarrow (7, 2500)$$

$$\therefore E(X_4 | 100, 950, 450) = \frac{\theta}{\alpha - 1} = \frac{2500}{6} = 416.67$$

Example 17.11 p. 413

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Generalize the result of Example 17.10 for an arbitrary sample size of n and an arbitrary prior gamma distribution with parameter α and β , where β is the reciprocal of the usual scale parameter.

Ans: The Bayesian premium estimate is given by

$$E(X_{n+1} | X=x) = \int \mu(\theta) \prod_{i=1}^n \theta^x e^{-\theta x} d\theta$$

* From previous Examples (17.3, 17.8 and 17.10), we can determine the posterior distⁿ as follows.

$$\prod_{i=1}^n \theta^x e^{-\theta x} \propto \left(\prod_{j=1}^n \theta e^{-\theta x_j} \right) \frac{\theta^{\alpha-1} e^{-\beta\theta} \beta^\alpha}{\Gamma(\alpha)}$$

product of exp. density values gamma density $f_0(\theta | \alpha, \beta)$

i.e. $\prod_{i=1}^n \theta^x e^{-\theta x} \propto \theta^{n+\alpha-1} e^{-(\sum x_j + \beta)\theta}$

So, we can write the posterior density f_0 as gamma density f_0 with parameters $n+\alpha$, $\frac{1}{\sum x_j + \beta}$, where $(\sum x_j + \beta)^{-1}$ is the scale parameter.

$$\therefore \prod_{i=1}^n \theta^x e^{-\theta x} = \frac{[(\sum x_j + \beta)\theta]^{n+\alpha} \exp[-(\sum x_j + \beta)\theta]}{\theta \Gamma(n+\alpha)}$$

The Bayes estimate of X_{n+1} is the expected value of θ^{-1} using the posterior distribution. It is

Remember
 $X \sim \text{gamma}(\alpha, \theta)$
 $\Rightarrow f(x) = \frac{(\theta^\alpha)^x e^{-x\theta}}{x \Gamma(\alpha)}$
 see p. 497.

$$E(X_{n+1} | X=x)$$

$$= \int \frac{1}{\theta} \frac{[(\sum x_j + \beta)\theta]^{n+\alpha} \exp[-(\sum x_j + \beta)\theta]}{\theta \Gamma(n+\alpha)} d\theta$$

$$= \frac{1}{\Gamma(n+\alpha)} \int \frac{1}{\theta^2} [(\sum x_j + \beta)\theta]^{n+\alpha} \exp[-(\sum x_j + \beta)\theta] d\theta$$

scale $\theta \rightarrow (\sum x_j + \beta)^{-1}$

let $(\sum x_j + \beta)\theta = u$

$$\therefore E(X_{n+1} | X=x) = \frac{(\sum x_j + \beta)^2}{\Gamma(n+\alpha)} \int \frac{u^{n+\alpha} e^{-u} du}{u^2 (\sum x_j + \beta)}$$

$$= \frac{\sum x_j + \beta}{\Gamma(n+\alpha)} \int u^{n+\alpha-2} e^{-u} du = \frac{(\sum x_j + \beta) \Gamma(n+\alpha-1)}{\Gamma(n+\alpha)}$$

$$\therefore E(X_{n+1} | X=x) = \frac{(\sum x_j + \beta)(n+\alpha-2)!}{(n+\alpha-1)!} = \frac{\sum x_j + \beta}{n+\alpha-1} = \frac{n \bar{x} + \alpha-1}{n+\alpha-1} \left(\frac{\beta}{\alpha-1} \right)$$



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Example 17.16 p. 421

Determine the Bühlmann estimate for the setting in Ex. (17.11)

Ans:

For the model discussed in EX. 17.11 p. 413

$$\mu(\theta) = \theta^{-1} \quad (1), \quad v(\theta) = \theta^{-2} \quad (2)$$

$$\begin{aligned} \mu = E(\theta^{-1}) &= \int_0^{\infty} \frac{1}{\theta} \frac{(\beta\theta)^{\alpha} e^{-\beta\theta}}{\Gamma(\alpha)} d\theta \\ &= \frac{\beta^{\alpha}}{\Gamma(\alpha)} \int_0^{\infty} \theta^{\alpha-2} e^{-\beta\theta} d\theta \end{aligned}$$

$$\mu = E(\theta^{-1}) = \frac{\beta^{\alpha}}{\Gamma(\alpha)} \int_0^{\infty} \left(\frac{u}{\beta}\right)^{\alpha-2} e^{-u} \frac{du}{\beta}, \quad u = \beta\theta$$

$$\therefore \mu = E(\theta^{-1}) = \frac{\beta}{\Gamma(\alpha)} \Gamma(\alpha-1) = \frac{\beta (\alpha-2)!}{(\alpha-1)!} = \frac{\beta}{\alpha-1} \quad (3)$$

$$\begin{aligned} v = E(\theta^{-2}) &= \int_0^{\infty} \frac{1}{\theta^2} \frac{(\beta\theta)^{\alpha} e^{-\beta\theta}}{\Gamma(\alpha)} d\theta \\ &= \frac{\beta^{\alpha}}{\Gamma(\alpha)} \int_0^{\infty} \theta^{\alpha-3} e^{-\beta\theta} d\theta \end{aligned}$$

$$\begin{aligned} &= \frac{\beta^{\alpha}}{\Gamma(\alpha)} \int_0^{\infty} \left(\frac{u}{\beta}\right)^{\alpha-3} e^{-u} \frac{du}{\beta} \\ &= \frac{\beta^2}{\Gamma(\alpha)} \int_0^{\infty} u^{\alpha-3} e^{-u} du \end{aligned}$$

$$v = E(\theta^{-2}) = \frac{\beta^2}{\Gamma(\alpha)} \Gamma(\alpha-2) = \frac{\beta^2 (\alpha-3)!}{(\alpha-1)!}$$

$$\therefore v = E(\theta^{-2}) = \frac{\beta^2}{(\alpha-1)(\alpha-2)} \quad (4)$$

$$a = \text{Var}(\theta^{-1}) = \frac{\beta^2}{(\alpha-1)(\alpha-2)} - \frac{\beta^2}{(\alpha-1)^2}$$

$$\therefore a = \frac{\beta^2}{\alpha-1} \left(\frac{1}{\alpha-2} - \frac{1}{\alpha-1} \right) = \frac{\beta^2}{(\alpha-1)^2 (\alpha-2)} \quad (5)$$

$\mu(\theta) = E(X_j | \theta = \theta)$
and $v(\theta) = \text{Var}(X_j | \theta = \theta)$
 $= \frac{1}{\theta}$ for exp. dist.
 $= \frac{1}{\theta^2}$ " " "

Remember
 $X \sim \text{gamma}(\alpha, \beta), \beta = \frac{1}{\theta}$
 $\Rightarrow f(x) = \frac{(\alpha\beta)^{\alpha} e^{-\beta x}}{\Gamma(\alpha)}$
Here, $x \rightarrow \theta \sim \text{gamma}(\alpha, \beta)$



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$$k = \frac{v}{a} = \frac{\beta^2}{(\alpha-1)(k-2)} \bigg/ \frac{\beta^2}{(\alpha-1)^2(\alpha-2)}$$

$$\therefore k = \alpha - 1 \quad (6)$$

The Bühlmann credibility factor is

$$Z = \frac{n}{n+k} = \frac{n}{n+\alpha-1} \quad (7)$$

∴ the Bühlmann credibility premium is given by

(3), (7) ⇒

$$P_c = Z\bar{X} + (1-Z)\mu$$

$$P_c = \frac{n}{n+\alpha-1} \bar{X} + \left(1 - \frac{n}{n+\alpha-1}\right) \frac{\beta}{\alpha-1}$$

$$\therefore P_c = \frac{n}{n+\alpha-1} \bar{X} + \frac{\alpha-1}{n+\alpha-1} \frac{\beta}{\alpha-1} \quad (8)$$

which again matches the Bayesian estimate that introduced in Ex. 17.11 p.413.

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