

EXAM C QUESTIONS OF THE WEEK

S. Broverman, 2006

Week of July 31/06

The following six losses were observed in 2005:

100 , 300 , 300 , 400 , 700 , 1000 .

The following three losses have been observed in 2006:

200 , 300 , 800 .

The model for the loss random variable in 2005 is exponential with a mean of θ .

It is assumed that there is 10% inflation in loss amounts from 2005 to 2006.

Two actuaries estimate θ using maximum likelihood estimation.

Actuary A uses an exponential distribution with mean 1.1θ as the model for the three losses in 2006 combined with the data for 2005 to estimate θ . The estimate is $\hat{\theta}_A$.

Actuary B also uses an exponential distribution with mean 1.1θ as the model for losses in 2006. Actuary B scales each of the losses for 2005 by 10% and uses those as equivalent losses in 2006 along with the actual three losses in 2006 to estimate θ . The estimate is $\hat{\theta}_B$.

Find $\hat{\theta}_B - \hat{\theta}_A$.

Solution can be found below.

Week of July 31/06 - Solution

The pdf of the 2005 loss random variable is $\frac{1}{\theta}e^{x/\theta}$, and the log is $-\ln \theta - \frac{x}{\theta}$.

The pdf of the 2006 loss random variable is $\frac{1}{1.1\theta}e^{x/1.1\theta}$, and the log is $-\ln 1.1\theta - \frac{x}{1.1\theta}$.

We denote losses as x_1, \dots, x_6 in 2005 and y_1, y_2, y_3 in 2006.

For Actuary A, the log-likelihood for the 2005 data is $-6 \ln \theta - \frac{\sum x_i}{\theta}$,

and for the 2006 data it is $-3 \ln 1.1\theta - \frac{\sum y_i}{1.1\theta}$.

Actuary A's total log-likelihood is

$$-6 \ln \theta - \frac{\sum x_i}{\theta} - 3 \ln 1.1\theta - \frac{\sum y_i}{1.1\theta} = -3 \ln 1.1 - 9 \ln \theta - \frac{\sum x_i}{\theta} - \frac{\sum y_i}{1.1\theta}.$$

Actuary B uses scaled losses $1.1x_1, \dots, 1.1x_6$ and losses y_1, y_2, y_3 for the exponential distribution with mean 1.1θ . The log-likelihood function for Actuary B is

$$-9 \ln 1.1\theta - \frac{1.1\sum x_i}{1.1\theta} - \frac{\sum y_i}{1.1\theta} = -9 \ln 1.1 - 9 \ln \theta - \frac{\sum x_i}{\theta} - \frac{\sum y_i}{1.1\theta}.$$

Since the two log-likelihood functions differ by a constant, they will both be maximized at the same value of θ . Therefore $\hat{\theta}_A = \hat{\theta}_B$.