

Exam in STK3505/4505: Answers

Problem 1

a

The moment estimates are given by:

$$\frac{\hat{\beta}}{\hat{\alpha} - 1} = \bar{z} \quad \text{and} \quad \bar{z} \sqrt{\frac{\hat{\alpha}}{\hat{\alpha} - 2}} = s,$$

such that

$$\hat{\alpha} = \frac{2}{1 - \bar{z}^2/s^2} = 4.01 \quad \text{and} \quad \hat{\beta} = \bar{z}(\hat{\alpha} - 1) = 8.04.$$

The variance is only defined when $\alpha > 2$. Therefore, moment estimation cannot be used for the Pareto distribution when $\alpha \leq 2$, and one may also run into numerical problems when α is > 2 , but close to 2, i.e. when the Pareto distribution is very heavy-tailed.

b

Simulation of Z using the inversion method requires an expression for the inverse cdf $F^{-1}(u)$:

$$F(z) = 1 - \frac{1}{\left(1 + \frac{z}{\beta}\right)^\alpha} = u$$
$$z = \beta \left((1 - u)^{-1/\alpha} - 1 \right) = F^{-1}(u).$$

Simulation algorithm:

- 1: Input: α, β
- 2: Draw $U^* \sim U(0, 1)$
- 3: Return $Z^* = \beta \left((1 - U^*)^{-1/\alpha} - 1 \right)$ % or $\beta \left((U^*)^{-1/\alpha} - 1 \right)$

c

Simulation algorithm for \mathcal{X} :

- 1: Input: $\lambda, \alpha, \beta, m$
- 2: **for** $i=1, \dots, m$ **do**
- 3: Draw $\mathcal{N}^* \sim \text{Poisson}(\lambda)$
- 4: $\mathcal{X}_i^* \leftarrow 0$
- 5: **for** $j=1, \dots, \mathcal{N}^*$ **do**

6: Draw $Z^* \sim \text{Pareto}(\alpha, \beta)$
7: $\mathcal{X}_i^* \leftarrow \mathcal{X}_i^* + Z^*$
8: **end for**
9: **end for**
10: Return $\mathcal{X}_1^*, \dots, \mathcal{X}_m^*$.

Estimate of the mean: $\bar{\mathcal{X}}^* = \frac{1}{m} \sum_{i=1}^m \mathcal{X}_i^*$.

Estimate of the standard deviation: $s_{\mathcal{X}}^* = \sqrt{\frac{1}{m-1} \sum_{i=1}^m (\mathcal{X}_i^* - \bar{\mathcal{X}}^*)^2}$.

The $100 \cdot \epsilon\%$ reserve q_ϵ is given by

$$P(\mathcal{X} \leq q_\epsilon) = \epsilon,$$

and estimated by $q_\epsilon^* = \mathcal{X}_{(m\epsilon)}^*$, where $\mathcal{X}_{(1)}^* \leq \dots \leq \mathcal{X}_{(m)}^*$.

d

$$E(\mathcal{X}) = \lambda \frac{\beta}{\alpha-1} = 80.1.$$

$$\text{sd}(\mathcal{X}) = \frac{\beta}{\alpha-1} \sqrt{\lambda \left(\frac{\alpha}{\alpha-2} + 1 \right)} = 25.3.$$

e

The 95% and 99% reserves are 125.5 and 153.0, respectively.

f

Simulations of cedent net portfolio payoffs $\mathcal{X}_1^{ce,*}, \dots, \mathcal{X}_m^{ce,*}$ are obtained by applying:

$$\mathcal{X}_i^{ce,*} = \mathcal{X}_i^* - \min(\max(\mathcal{X}_i^* - a, 0), b)$$

to the simulations from **c**. The corresponding $100 \cdot \epsilon\%$ cedent net reserve is estimated by $q_\epsilon^{ce,*} = \mathcal{X}_{(m\epsilon)}^{ce,*}$, where $\mathcal{X}_{(1)}^{ce,*} \leq \dots \leq \mathcal{X}_{(m)}^{ce,*}$.

g

The cedent net 95% and 99% reserves are 90.3 and 118.4, respectively. Obviously, the reserves become lower when part of the responsibility is transferred on a reinsurer. Since there is an upper bound on this responsibility which is close to the 95% reserve without reinsurance, the reduction in the 99% reserve is smaller than that of the 95% reserve. The reinsurance pure premium is $\pi^{re,pu} = E(\mathcal{X}^{re}) = E(\mathcal{X}) - E(\mathcal{X}^{ce}) = 5.2$ and the actual premium is $\pi^{re} = (1 + \gamma)\pi^{re,pu} = 9.36$.

Problem 2

a

$$\pi_{l_0} = \begin{cases} s \sum_{k=l_r-l_0}^{\infty} d^k {}_k p_{l_0}, & l_0 < l_r \\ s \sum_{k=0}^{\infty} d^k {}_k p_{l_0}, & l_0 \geq l_r \end{cases},$$

where ${}_k p_{l_0}$ takes the risk of death into account, whereas d^k is the discount factor.

b

The shorter the time between the start age l_0 and the age of retirement l_r , the less the pension is discounted. Further, the probability of surviving til and past the age of retirement increases with increasing age. Hence, the present value increases with l_0 . This explains the upwards shape of the curves in the plot. Further, a higher value of the discount rate r leads to a heavier discount of the pension payments in the future, i.e. smaller discount factors d^k , and hence a smaller present value of the pension. This means that the solid, dashed, dotted and dash-dotted curves correspond to $r = 0.05, 0.04, 0.03$ and 0.02 , respectively.

c

Present value of payments: $\zeta \sum_{k=0}^{l_r-l_0-1} d^k {}_k p_{l_0}$, where as in **a** ${}_k p_{l_0}$ takes the risk of death into account, whereas d^k is the discount factor.

d

Equivalence means that the expected present value of the payments ζ should be equal to the expected present value of the pension π_{l_0} , so that:

$$\zeta = s \frac{\sum_{k=l_r-l_0}^{\infty} d^k {}_k p_{l_0}}{\sum_{k=0}^{l_r-l_0-1} d^k {}_k p_{l_0}}.$$

Problem 3

a

$$R_k = \frac{S_k}{S_{k-1}} - 1 = \frac{e^{Y_k}}{e^{Y_{k-1}}} - 1 = e^{Y_k - Y_{k-1}} - 1 = e^{X_k} - 1.$$

$$\text{K-step return: } R_{0:K} = (1 + R_1) \cdot \dots \cdot (1 + R_K) - 1.$$

$$R_{0:K} = (1 + R_1) \cdot \dots \cdot (1 + R_K) - 1 = e^{X_1} \cdot \dots \cdot e^{X_K} - 1 = e^{\sum_{j=1}^K X_j} - 1 = e^{K\xi + \sigma \sum_{j=1}^K \varepsilon_j}.$$

Let $\eta = \frac{1}{\sqrt{K}} \sum_{j=1}^K \varepsilon_j$. Since $\varepsilon_j \stackrel{iid}{\sim} N(0, 1)$, η is also normally distributed with

$$\begin{aligned} \mathbb{E}(\eta) &= \frac{1}{\sqrt{K}} \sum_{j=1}^K \mathbb{E}(\varepsilon_j) = 0 \\ \text{Var}(\eta) &= \frac{1}{K} \sum_{j=1}^K \text{Var}(\varepsilon_j) = 1. \end{aligned}$$

Thus, $R_{0:K} = e^{K\xi + \sigma\sqrt{K}\eta} - 1$, with $\eta \sim N(0, 1)$.

b

Simulation algorithm for X :

- 1: Input: r, σ, v_0, r_g, K
- 2: $\begin{cases} R^* \leftarrow 0 \\ P^* \leftarrow 1 \end{cases}$
- 3: **for** $k=1, \dots, K$ **do**
- 4: Draw $\varepsilon^* \sim N(0, 1)$
- 5: $R^* \leftarrow e^{\xi + \sigma\varepsilon^*} - 1$
- 6: $P^* \leftarrow P^*(1 + R^*)$
- 7: **end for**
- 8: Return $R_{0:K}^* \leftarrow P^* - 1$.

With m simulations $R_{0:K,1}^*, \dots, R_{0:K,m}^*$ from the above algorithm, the pdf of $R_{0:K}$ can be estimated by

$$f^*(R) = \frac{1}{mh} \sum_{i=1}^m K \left(\frac{R - R_{0:K,i}^*}{h} \right),$$

where $K(\cdot)$ is a kernel function with mean 0 that integrates to 1, for instance a Gaussian kernel. This can be seen as a smoothing of a histogram, where the degree of smoothness is controlled by the bandwidth parameter h . Small values of h give a low bias, but a high variance and a un-smooth estimate. Higher values of h give a higher bias, but also a lower variance and a smoother estimate. The choice of bandwidth should therefore be a trade-off between bias and variance.

c As the ε_k s follow a normal distribution, so does $X_k|x_0$. This means that $e^{X_k}|x_0$ follows a log-normal distribution with parameters $\left(a^k x_0, \sqrt{\frac{1-a^{2k}}{1-a^2}} \tau \right)$,

and its mean and standard deviation are as given in the formulas on the first page of the exam. We have:

$$\begin{aligned}
 \mathbb{E}(r_k|r_0) &= r e^{-\tau^2/(2(1-a^2))} \mathbb{E}(e^{X_k}|x_0) \\
 &= r e^{-\tau^2/(2(1-a^2))} e^{a^k x_0 + \frac{1-a^{2k}}{2(1-a^2)} \tau^2} \\
 &= r e^{a^k x_0 - \tau^2 a^{2k}/(2(1-a^2))}
 \end{aligned}$$

and

$$\begin{aligned}
 \text{sd}(r_k|r_0) &= r e^{-\tau^2/(2(1-a^2))} \text{sd}(e^{X_k}|x_0) \\
 &= r e^{-\tau^2/(2(1-a^2))} e^{a^k x_0 + \frac{1-a^{2k}}{2(1-a^2)} \tau^2} \sqrt{e^{\frac{1-a^{2k}}{1-a^2} \tau^2} - 1} \\
 &= \mathbb{E}(r_k|r_0) \sqrt{e^{\tau^2(1-a^{2k})/(1-a^2)} - 1}.
 \end{aligned}$$

d The stock investment gives a high expected return and has a very high upside, with a 95% quantile of 5.50. The bank account with a floating interest rate obviously gives a much lower expected return and upside. On the other hand, the risk is also much lower, which is seen both on the standard deviation, that is much lower, and the 5% quantile, that is much higher than for the stock return. The choice of investment must therefore depend on the risk appetite of the investor.