## Problem 1

$$
\mathbf{a})\ \pi^{pu} = \mathrm{E}(X).
$$

b)  $\pi = (1 + \gamma)\pi^{pu}$ , where  $\pi$  is the actual premium payed and  $\gamma$  the loading

c) Let  $\mathcal{X} = X_1 + \ldots + X_J$  be the total payout from the portfolio. The 100 $\cdot \epsilon$ % reserve  $q_{\epsilon}$  is given by

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

where  $\epsilon$  is close to 1.

d)  $E(X_1 + ... + X_J) = \sum_{j=1}^{J} E(X_j) = J\xi$  and  $Var(X_1 + ... + X_J)$  indep- $\sum_{j=1}^{J} \text{Var}(X_j) = J\sigma^2$ . Hence,

$$
\frac{\mathrm{sd}(X_1 + \ldots + X_J)}{\mathrm{E}(X_1 + \ldots + X_J)} = \frac{\sqrt{J}\sigma}{J\xi} = \frac{\sigma/\xi}{\sqrt{J}} \underset{J \to \infty}{\longrightarrow} 0.
$$

# Problem 2

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

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

$$
(1 + z/\beta)^{\alpha} = (1 - u)^{-1}
$$
  

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

Simulation algorithm:

- 1: Input:  $\alpha, \beta$
- 2: Draw  $U^* \sim U(0, 1)$
- 3: Return  $Z^* = \beta ((1 U^*)^{-1/\alpha} 1)$  % or  $\beta ((U^*)^{-1/\alpha} 1)$
- **b**) Simuleringsalgoritme for  $\mathcal{X}$ :

1: Input: 
$$
\lambda, \alpha, \beta, m
$$

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

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

$$
E(\mathcal{X}) = E(E(\mathcal{X}|\mathcal{N}))
$$
  
= 
$$
E\left(E\left(\sum_{i=1}^{N} Z_i|\mathcal{N}\right)\right)
$$
  
= 
$$
E(\mathcal{N}E(Z_i))
$$
  
= 
$$
E(\mathcal{N})E(Z_i)
$$
  
= 
$$
\lambda \frac{\beta}{\alpha - 1} = 25.
$$

d) The 95% and 99% reserves are 48.8 and 64.2 respectively.

e) Simulations of reinsurer portfolio payoffs  $\mathcal{X}_1^{re,*}$  $\mathcal{X}_1^{re,*}, \ldots, \mathcal{X}_m^{re,*}$  are obtained by applying  $\mathcal{X}_i^{re,*} \leftarrow \max(\mathcal{X}_i^* - a, 0)$  to the simulations from b).

f) The pure reinsurance premium is 3.15 and the reinsurance 95% and 99% reserves are 18.76 and 34.25 respectively.

#### Problem 3

a) Let Y be the number of years an individual is alive.

$$
k p_{l_0} = P(Y > l_0 + k | Y > l_0)
$$
  
= 
$$
\frac{P(Y > l_0 + k)}{P(Y > l_0)}
$$
  
= 
$$
\frac{P(Y > l_0 + k)}{P(Y > l_0 + k - 1)} \cdot \frac{P(Y > l_0 + k - 1)}{P(Y > l_0 + k - 2)} \cdot \dots \cdot \frac{P(Y > l_0 + 1)}{P(Y > l_0)}
$$
  
= 
$$
P(Y > l_0 + k | Y > l_0 + k - 1) \cdot \dots \cdot P(Y > l_0 + 1 | Y > l_0)
$$
  
= 
$$
p_{l_{0+k-1}} p_{l_{0+k-2}} \cdot \dots \cdot p_{l_0}
$$

**b)** 
$$
V_0 = -\pi \sum_{i=0}^{K-1} d^i \, i p_{l_0} + s \sum_{i=K}^{\infty} d^i \, i p_{l_0}.
$$
  
\n**c)**  $\pi = s \frac{\sum_{i=K}^{\infty} d^i \, i p_{l_0}}{\sum_{i=0}^{K-1} d^i \, i p_{l_0}}$   
\n**d)**  $V_k = \begin{cases} -\pi \sum_{i=k}^{K-1} d^{i-k} \, i p_{l_0} + s \sum_{i=K}^{\infty} d^{i-k} \, i p_{l_0}, & k < K \\ s \sum_{i=k}^{\infty} d^{i-k} \, i p_{l_0}, & k \ge K \end{cases}$ 

e) The sum to be repaid is  $V_k$ .

# Problem 4

a)  $R_k = \frac{S_k}{S_k}$  $\frac{S_k}{S_{k-1}} - 1 = \frac{e^{Y_k}}{e^{Y_{k-1}}}$  $\frac{e^{Y_k}}{e^{Y_{k-1}}} - 1 = e^{Y_k - Y_{k-1}} - 1 = e^{X_k} - 1.$ **b**)  $Y_k = Y_{k-1} + X_k = Y_{k-2} + X_{k-1} + X_k = \ldots = Y_0 + \sum_{j=1}^k X_j = \log(s_0) +$  $\sum_{j=1}^k X_j$ . Since  $X_j \stackrel{iid}{\sim} N(\xi, \sigma)$ ,  $Y_k$  is also normally distributed with

$$
E(Y_k) = \log(s_0) + E\left(\sum_{j=1}^k X_j\right) = \log(s_0) + \sum_{j=1}^k E(X_j) = \log(s_0) + k\xi
$$
  

$$
sd(Y_k) = sd\left(\sum_{j=1}^k X_j\right)^{indep.} \sqrt{\sum_{j=1}^k Var(X_j)} = \sqrt{k}\sigma.
$$

Then,  $S_k = e^{Y_k}$  must follow a log-normal distribution with

$$
E(S_k) = e^{\log(s_0) + k\xi + \frac{1}{2}k\sigma^2} = s_0 e^{k\xi + \frac{1}{2}k\sigma^2}.
$$

c) Simulation algorithm for  $S_1, \ldots, S_K$ :

1: Input:  $\xi, \sigma, s_0, K$ 2:  $Y^* \leftarrow \log(s_0)$ 3: for  $k=1,\ldots,K$  do 4: Draw  $X^* \sim N(\xi, \sigma)$ 5:  $Y^* \leftarrow Y^* + X^*$ 6:  $S^* \leftarrow e^{Y^*}$ 7: end for 8: Return  $S_1^*, \ldots, S_K^*$ .

**d)**  $E(S_k) = s_0 e^{k\xi + \frac{1}{2}k\sigma^2} \approx 2.01.$ 

e)  $R_{0:K} = (1 + R_1) \cdot \ldots \cdot (1 + R_K) - 1 = e^{X_1} \cdot \ldots \cdot e^{X_K} - 1 = e^{\sum_{j=1}^K X_j} - 1 =$  $e^{Y_K-\log(s_0)}-1=\frac{S_K}{s_0}-1$ . Hence,  $P(R_{0:K}\leq r)=P\left(\frac{S_K}{s_0}\right)$  $\frac{S_K}{s_0} - 1 \leq r$ ) = P( $S_K \leq$  $s_0(r+1)$ .

- Probability of losing on the investment:  $P(R_{0:K} < 0) = P(S_K < s_0)$  $P(S_K < 1)$ , which is a little smaller than 0.25.
- Probability of doubling the investment:  $P(R_{0:K} \geq 1) = P(S_K \geq 2s_0)$  $1 - P(S_K < 2)$ , which is between 0.25 and 0.5.
- Probability of quadruplicating the investement:  $P(R_{0:K} \geq 3) = P(S_K \geq 1)$  $(4s_0) = 1 - P(S_K < 4)$ , which is between 0.05 and and 0.25, but closer to 0.05.

### Problem 5

a) Simulation algorithm for  $B_1, \ldots, B_K$ :

1: Input: 
$$
\xi, a, \tau, b_0, r_0, K
$$
  
\n2: 
$$
\begin{cases} B_0^* \leftarrow b_0 \\ Z^* \leftarrow \log \left( \frac{r_0}{\xi} \right) \\ 3: \text{ for } k=1,\ldots,K \text{ do} \\ 4: \quad \text{Draw } \varepsilon^* \sim N(0,1) \\ 5: \quad \begin{cases} Z^* \leftarrow aZ^* + \tau \varepsilon^* \\ r^* \leftarrow \xi e^{Z^*} \\ 6: \quad B_k^* \leftarrow (1+r^*)B_{k-1}^* \\ 7: \text{ end for} \\ 8: \text{ Return } B_1^*, \ldots, B_K^*. \end{cases}
$$

b) Since  $S_k$  and  $B_k$  are independent, they can be simulated independently from the algorithms in 4 c) and 5 a), respectively, resulting in  $S_1^*, \ldots, S_m^*$  and  $\bm{B}_{1}^{*}, \ldots, \bm{B}_{m}^{*}$ , with  $\bm{S}_{j}^{*} = (S_{1j}^{*}, \ldots, S_{Kj}^{*})^{T}$  and  $\bm{B}_{j}^{*} = (B_{1j}^{*}, \ldots, B_{Kj}^{*})^{T}$ . Then, we obtain  $V_1^*, \ldots, V_m^*$ , with  $V_j^* = (V_{1j}^*, \ldots, V_{Kj}^*)^T$ , from  $V_j^* = S_j^* + B_j^*$ .

c) When half of the investment is placed in a cash account instead of stocks, the risk becomes lower, which is seen in the higher values for left tail quantiles and  $\frac{\text{sd}(V_K)}{\text{E}(V_K)} \approx 0.41 < \frac{\text{sd}(S_K)}{\text{E}(S_K)}$  $\frac{\text{SU}(S_K)}{\text{E}(S_K)} \approx 0.7$ . On the other hand, the upside is also lower, which is seen in the lower expectation and right tail quantiles.