

**Risk theory and risk management in actuarial science**  
**Winter term 2020/2021**

**6th work sheet**

**37. Equivalent threshold models**

Let  $X = (X_1, X_2, \dots, X_m)'$  be an  $m$ -dimensional random vector and let  $D \in \mathbb{R}^{m \times n}$  be a deterministic matrix with elements  $d_{ij}$  such that for every  $i$ ,  $1 \leq i \leq m$ , the elements of the  $i$ -th row form a set of increasing thresholds satisfying  $d_{i,1} < d_{i,2} \dots < d_{i,n}$ . Introduce additionally  $d_{i0} = -\infty$ ,  $d_{i,n+1} = +\infty$  and set

$$S_i = j \iff d_{ij} < X_i \leq d_{i,j+1}, \text{ for } j \in \{0, \dots, n\}, i \in \{1, \dots, m\}.$$

Then  $(X, D)$  is said to define a *threshold model for the state vector*  $S = (S_1, \dots, S_m)'$ . We refer to  $X$  as the *vector of critical variables* and denote its marginal distribution functions by  $F_i(x) = P(X_i \leq x)$ , for  $i \in \{1, 2, \dots, m\}$ . The  $i$ -th row of  $D$  contains the critical thresholds for firm  $i$ . By definition, default (corresponding to event  $S_i = 0$ ) occurs iff  $X_i \leq d_{i1}$ , thus the default probability of company  $i$  is given by  $\bar{p}_i := F_i(d_{i1})$ . Let  $Y_i$  be the default indicator of company  $i$ , i.e.  $Y_i \in \{0, 1\}$  with  $Y_i = 1$  iff company  $i$  defaults, hence  $Prob(Y_i = 1) = \bar{p}_i$  and  $Prob(Y_i = 0) = 1 - \bar{p}_i$ , for  $1 \leq i \leq m$ . We denote by  $\rho(Y_i, Y_j)$  the *default correlation* of two firms  $i \neq j$ ; this quantity depends on  $E(Y_i, Y_j)$  (how?) which in turn depends on the joint distribution of  $(X_i, X_j)$ , and hence on the copula of  $(X_1, X_2, \dots, X_m)'$ . (Notice that in general the latter is not fully determined by the *asset correlation*  $\rho(X_i, X_j)$ .)

Two threshold models  $(X, D)$  and  $(\tilde{X}, \tilde{D})$  for the state vectors  $S$  and  $\tilde{S}$ , respectively, are called equivalent, iff  $S$  and  $\tilde{S}$  have the same probability distribution.

Show that two threshold models  $(X, D)$  and  $(\tilde{X}, \tilde{D})$  with state vectors  $S$  and  $\tilde{S}$ , respectively, are equivalent if the following conditions hold:

- (a) The marginal distributions of the random vectors  $S$  and  $\tilde{S}$  coincide, i.e.  $P(S_i = j) = P(\tilde{S}_i = j)$ , for all  $j \in \{1, \dots, n\}$ ,  $i \in \{1, \dots, m\}$ .
- (b)  $X$  and  $\tilde{X}$  admit the same copula  $C$ .

**38. A bank has a loan portfolio of 100 loans. Let  $X_k$  be the default indicator for loan  $k$  such that  $X_k = 1$  in case of default and 0 otherwise, for  $k \in \{1, \dots, 100\}$ .**

- (a) Suppose that  $X_k$  are independent and identically distributed with  $P(X_k = 1) = 0.01$ . Compute the expected value  $E(N)$  of the number  $N$  of defaults and  $P(N = k)$  for  $k \in \{0, 1, \dots, 100\}$ .
- (b) Consider the risk factor  $Z$  which reflects the state of the economy. Suppose that conditional on  $Z$  the default indicators are independent and identically distributed with  $P(X_k = 1|Z) = Z$ , where  $P(Z = 0.01) = 0.9$  and  $P(Z = 0.11) = 0.1$ . Compute the expected value  $E(N)$  where  $N$  is defined as in (a).
- (c) Consider the risk factor  $Z$  which reflects the state of the economy. Suppose that conditional on  $Z$  the default indicators are independent and identically distributed with  $P(X_k = 1|Z) = Z^9$ , where  $Z$  is uniformly distributed on  $(0, 1)$ . Compute the expected value  $E(N)$  where  $N$  is defined as in (a).

**39. An  $m$ -dimensional random vector  $X$  is said to have a  $p$ -dimensional conditional independence structure with conditioning variable  $\Psi$  iff there is some  $p \in \mathbb{N}$ ,  $p < m$ , and a  $p$ -dimensional random vector  $\Psi = (\Psi_1, \dots, \Psi_p)'$ , such that the random variables  $X_1, \dots, X_m$  are independent conditional on  $\Psi$ . Consider a threshold model  $(X, D)$  as defined in Exercise 37 and assume that  $X$  has a  $p$ -dimensional conditional independence structure with conditioning variable  $\Psi$ . Show that the default indicators  $Y_i = \mathbb{I}_{X_i \leq d_{i1}}$  follow a Bernoulli mixture model with factor  $\Psi$ . How are given the conditional default probabilities for this Bernoulli mixture model?**

40. Suppose that the critical variables  $X = (X_1, \dots, X_m)^t$  have a normal mean-variance mixture distribution, i.e.  $X = m(W) + \sqrt{W}Z$  with an  $m$ -dimensional random vector  $Z$ , a positive, scalar random variable  $W$  independent of  $Z$ , and a measurable function  $m: [0; +\infty) \rightarrow \mathbb{R}^m$ . Assume that  $Z$  (and hence  $X$ ) follows a linear factor model of the form  $Z = BF + \epsilon$ , where  $F \sim N_p(\vec{0}, \Omega)$  is a  $p$ -dimensional normally distributed vector with expected value  $\vec{0} \in \mathbb{R}^p$  and covariance matrix  $\Omega \in \mathbb{R}^{p \times p}$ ,  $B \in \mathbb{R}^{m \times p}$  is a deterministic loading matrix, and the components  $\epsilon_1, \dots, \epsilon_m$  of  $\epsilon$  are i.i.d. normally distributed random variables which are also independent of  $F$ . Show that  $F$  has a  $(p+1)$ -conditional independence structure (see the definition in Exercise 39). How are given the conditional default probabilities for the corresponding Bernoulli mixture model of the default indicators  $Y_i$  in this case (cf. Exercise 39)?

41. (Application of Archimedean copulas in threshold models)

Consider a threshold model  $(X, D)$  where  $X$  has an Archimedean copula  $C$  with generator  $\phi$  such that  $\phi^{-1}$  is the Laplace transform of some nonnegative distribution function  $G$  with  $G(0) = 0$ . Let  $d = (d_{11}, \dots, d_{m1})^t$  denote the first column of  $D$  containing the default thresholds. We write then  $(X, d)$  for a threshold model of default with an Archimedean copula dependence and denote by  $\bar{p} = (\bar{p}_1, \dots, \bar{p}_m)^t$  the vector of default probabilities, where  $\bar{p}_i = \mathbb{P}(X_i \leq d_{i1})$ , for  $i \in \{1, 2, \dots, m\}$ . Consider a nonnegative random variable  $\Psi \sim G$  and random variables  $U_1, \dots, U_m$  that are conditionally independent given  $\Psi$  with conditional distribution function  $\mathbb{P}(U_i \leq u | \Psi = \psi) = \exp(-\psi\phi(u))$ , for  $u \in [0, 1]$ . Check that  $U = (U_1, \dots, U_m)^t$  has distribution function  $C$ . Show that  $(X, d)$  and  $(U, \bar{p})$  are two equivalent threshold models for default (cf. Exercise 37). How are given the conditional default probabilities  $p_i(\Psi)$  in this case?

Consider the Clayton copula  $C_\theta^{Cl}$  with generator function  $\phi(t) = t^{-\theta} - 1$  and assume that we want to construct a Bernoulli mixture model that is equivalent to a threshold model driven by  $C_\theta^{Cl}$ . In this Bernoulli mixture model all conditional default probabilities  $p_i(\Psi)$  would coincide; such a Bernoulli mixture model is called *exchangeable*. Assume moreover that the probability of default for any creditor is given by  $\pi$  and the probability that an arbitrary pair of creditors defaults is given by  $\pi'$ . What value of  $\theta$  would lead to the required exchangeable Bernoulli mixture model?

42. Apply the CreditRisk<sup>+</sup> approach for a credit portfolio with  $n = 100$  credits, and  $m$  risk factors, where  $m = 1$  or  $m = 5$ . Consider the settings  $\bar{\lambda}_i = \bar{\lambda} = 0.15$ ,  $\alpha_j = \alpha = 1$ ,  $\beta_j = \beta = 1$ ,  $a_{i,j} = 1/m$ , for  $i = 1, 2, \dots, n$ ,  $j = 1, 2, \dots, m$ . Let  $N$  be the number of defaulting creditors. Recall that  $\mathbb{P}(N = k) = \frac{1}{k!} g_N^{(k)}(0) = \frac{1}{k!} \frac{d^k g_N(0)}{dt^k}$  holds for any  $k \in \overline{1..n}$ , where  $g_N$  is the probability generating function (pgf) of  $N$  (cf. the lecture).

(a) Based on the closed form expressions discussed in the lecture derive concrete formulas for the probability generating functions  $\tilde{g}_N(t)$  and  $\bar{g}_N(t)$  of  $N$  in the cases  $m = 1$  and  $m = 5$ , respectively.

(b) Show that the following recursive formula holds for  $g_N = \tilde{g}_N$  and  $g = \bar{g}_N$  and  $k > 1$ :

$$g_N^{(k)}(0) = \sum_{l=0}^{k-1} \binom{k-1}{l} g_N^{(k-1-l)}(0) \sum_{j=1}^m l! \alpha_j \delta_j^{l+1}.$$

(c) Compute and plot  $\mathbb{P}(N = k)$ , for  $k \in \overline{1..n}$ , in both cases  $m = 1$  and  $m = 5$ . Compare the two plots and interpret the results.