# Risk theory and risk management in actuarial science Winter term 2017/18 

## 4th work sheet

21. Let the random variables $X_{i}, i=1,2$, be such that $X_{1} \sim \operatorname{Exp}(\lambda)$ and $X_{2}=X_{1}$, where $\operatorname{Exp}(\lambda)$ is the exponential distribution with parameter $\lambda$. Consider the strictly increasing functions $t_{i}: \mathbb{R} \rightarrow \mathbb{R}$, $i=1,2$, with $t_{1}(x)=x$ and $t_{2}(x)=x^{2}$. Show that the following equalities for the linear correlation coefficient $\rho_{L}$ hold:

$$
\rho_{L}\left(X_{1}, X_{2}\right)=1 \text { and } \rho_{L}\left(t_{1}\left(X_{1}\right), t_{2}\left(X_{2}\right)\right)=\frac{2}{\sqrt{5}}
$$

22. Let the random variables $X_{i}, i=1,2$, be such that $X_{1} \sim \operatorname{Exp}(\lambda)$ and $X_{2}=X_{1}^{2}$, where $\operatorname{Exp}(\lambda)$ is the exponential distribution with parameter $\lambda$. Determine the coefficients of the lower and the upper tail dependence $\lambda_{L}\left(X_{1}, X_{2}\right), \lambda_{U}\left(X_{1}, X_{2}\right)$, respectively, and conclude that $X_{1}$ and $X_{2}$ have both a lower and an upper tail dependence. Compute also the coefficient of the linear correlation $\rho_{L}\left(X_{1}, X_{2}\right)$, compare the three computed dependence measures and comment on your results.
23. (Properties of the generalized inverse function)

Let $h: \mathbb{R} \rightarrow \mathbb{R}$ be a monotone increasing function with $h(\mathbb{R})=\mathbb{R}$ and let $h^{\leftarrow}: \mathbb{R} \rightarrow \mathbb{R}$ be generalized inverse function of $h$. Show that the following statements hold.
(a) $h^{\leftarrow}$ is a monotone increasing left-continuous function.
(b) $h$ is continuous if and only if $h^{\leftarrow}$ is strictly monotone increasing .
(c) $h$ is strictly monotone increasing if and only if $h \leftarrow$ is continuous.
(d) $h \leftarrow(h(x)) \leq x$
(e) If $h$ is strictly monotone increasing then $h^{\leftarrow}(h(x))=x$.
(f) $h$ is continuous if and only if $h\left(h^{\leftarrow}(y)\right)=y$.
24. (A coherent premium principle)

Consider two constants $p>1$ and $\alpha \in[0,1)$. Let $(\Omega, \mathcal{F}, P)$ be some fixed probability space and $\mathcal{M}$ be the set of all random variables $L$ on $(\Omega, \mathcal{F})$ for which $E\left(|L|^{p}\right)^{1 / p}$ is finite, i.e. $E\left(|L|^{p}\right)^{1 / p}<\infty$. Define a risk measure $\rho_{\alpha, p}:=E(L)+\alpha\left(\left\|(L-E(L))^{+}\right\|_{p}\right.$ on $\mathcal{M}$, where $\|X\|_{p}:=E\left(|X|^{p}\right)^{1 / p}$ is the $L^{p}$-norm of the positive part of the centered random variable $X-E(X)$ for any random variable $X \in \mathcal{M}$. Show that $\rho_{\alpha, p}$ is a coherent risk measure for any $p>1$ and any $\alpha \in[0,1)$. So we get a whole family of coherent risk measures $\rho_{\alpha, p}$ for $p>1$ and $\alpha \in[0,1)$. How do the parameters $\alpha$ and $p$ influence $\rho_{\alpha, p}$ ? Which parameter values lead to more "conservative" risk measures?
25. (Generalized scenarios as coherent risk measures)

Denote by $\mathcal{P}$ a set of probability measures on some underlying measurable space $(\Omega, \mathcal{F})$ and set

$$
\mathcal{M}_{\mathcal{P}}:=\left\{L: L \text { is a r.v. on }(\Omega, \mathcal{F}), E^{Q}(|L|)<\infty \text { for all } Q \in \mathcal{P}\right\}
$$

where $E^{Q}(X)$ denotes the expected value of a random variable $X$ under the probability measure $Q$. Then the risk measure induced by the set of generalized scenarios $\mathcal{P}$ is the mapping $\rho_{\mathcal{P}}: \mathcal{M}_{\mathcal{P}} \rightarrow \mathbb{R}$ such that $\rho_{\mathcal{P}}(L):=\sup \left\{E^{Q}(L): Q \in \mathcal{P}\right\}$. Show that $\rho_{\mathcal{P}}$ is coherent on $\mathcal{M}_{\mathcal{P}}$ for any set $\mathcal{P}$ of probability measures on $\mathcal{M}_{\mathcal{P}}$. Interprete the scenario based risk measures (cf. lecture) as a risk measure generalized by an appropriately defined set of probability measures on appropriately defined discrete probability spaces ${ }^{1}$.

[^0]26. Show that the Fréchet lower bound $W_{d}$ is not a copula for $d \geq 3$.

Hint: Show that the rectangle inequality

$$
\sum_{k_{1}=1}^{2} \sum_{k_{2}=1}^{2} \ldots \sum_{k_{d}=1}^{2}(-1)^{k_{1}+k_{2}+\ldots+k_{d}} W_{d}\left(u_{1 k_{1}}, u_{2 k_{2}}, \ldots, u_{d k_{d}}\right) \geq 0
$$

where $\left(a_{1}, a_{2}, \ldots, a_{d}\right),\left(b_{1}, b_{2}, \ldots, b_{d}\right) \in[0,1]^{d}$ with $a_{k} \leq b_{k}$ and $u_{k 1}=a_{k}$ und $u_{k 2}=b_{k}$ for all $k \in\{1,2, \ldots, d\}$, is violated if $d \geq 3$ and $a_{i}=\frac{1}{2}, b_{i}=1$, for $i=1,2, \ldots, d$.
27. Let $X_{i}, i=1,2$, be two lognormally distributed random variables with $X_{1} \sim \operatorname{Lognormal}(0,1)$ und $X_{2} \sim \operatorname{Lognormal}\left(0, \sigma^{2}\right), \sigma>0$. Compute $\rho_{L, \min }\left(X_{1}, X_{2}\right)$ and $\rho_{L, \max }\left(X_{1}, X_{2}\right)$ in dependence of $\sigma$ and compare their values for different values of $\sigma>0$. What can you say about the copula of $\left(X_{1}, X_{2}\right)$ in each of the cases? Plot the graphs of $\rho_{L, \min }\left(X_{1}, X_{2}\right)$ and $\rho_{L, \max }\left(X_{1}, X_{2}\right)$ as functions of $\sigma$ and comment on the behaviour of these functions for $\sigma \rightarrow+\infty$ ?
Hint: Consider $X_{1}=\exp (Z)$ and $X_{2}=\exp (\sigma Z)$ or $X_{2}=\exp (-\sigma Z)$ for a standard normally distributed random variable $Z$.


[^0]:    ${ }^{1}$ It can be shown that in the case of discrete probability spaces any coherent risk measure is induced by some set of generalized scenarios as described above, see Proposition 6.11 in A.J. McNeil, R. Frey and P. Embrechts, Quantitative Risk Management: Concepts, Techniques amd Tools, Princeton Unversity Press, 2005.

