

A note on a class of certain mixtures of product –gamma distributions

Dietmar Pfeifer
Institut für Mathematik
Fakultät V
Carl-von-Ossietzky Universität Oldenburg

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In this note, we investigate multivariate distributions with densities of the form

$$h(x_1, \dots, x_n) = \left(\frac{1}{n} \sum_{k=1}^n x_k \right) \cdot \exp \left(- \sum_{k=1}^n x_k \right), \quad x_1, \dots, x_n \geq 0 \text{ with } n \in \mathbb{N}.$$
 Note that this density can be rewritten as

$$h(x_1, \dots, x_n) = \frac{1}{n} \sum_{k=1}^n x_k \cdot \exp(-x_k) \cdot \prod_{\substack{i=1 \\ i \neq k}}^n \exp(-x_i), \quad x_1, \dots, x_n \geq 0,$$
 i.e. it is the density of a mixture

of n product-gamma distributions where exactly one component is representing a $\Gamma(2,1)$ -distribution and all others represent independent $\mathcal{E}(1)$ -Exponential distributions. Note that an $\mathcal{E}(1)$ -Exponential distribution is identical to a $\Gamma(1,1)$ -distribution (see [1], p. 337, (17.2)). This implies that we can write the corresponding c.d.f. H as follows:

$$H(x_1, \dots, x_n) = \frac{1}{n} \sum_{(i_1, \dots, i_n) \in \mathcal{J}} \prod_{j=1}^n F_{i_j}(x_j) \text{ with } \mathcal{J} = \left\{ (i_1, \dots, i_n) \in \{0,1\}^n \mid \sum_{j=1}^n i_j = 1 \right\},$$
 i.e. $\mathcal{J}$ contains

all $\{0,1\}$ -valued n -tuples with all entries equal to zero except for exactly one entry equal to 1, and $F_0(x) = 1 - \exp(-x)$ and $F_1(x) = 1 - (1+x)\exp(-x)$, $x \geq 0$. The one-dimensional

marginal c.d.f. H_1 is thus given by $H_1(x) = \frac{F_1(x) + (n-1)F_0(x)}{n} = 1 - \frac{n+x}{n} \exp(-x)$ with

density $h_1(x) = \frac{f_1(x) + (n-1)f_0(x)}{n} = \left(1 + \frac{x-1}{n} \right) \exp(-x)$, $x \geq 0$ where $f_0(x) = \exp(-x)$ and

$f_1(x) = x \exp(-x)$, $x \geq 0$ while the two-dimensional marginal c.d.f. H_2 is correspondingly given by

$$H_2(x, y) = \frac{F_0(x)F_1(y) + F_1(x)F_0(y) + (n-2)F_0(x)F_0(y)}{n} \text{ with density}$$

D. Pfeifer (✉)

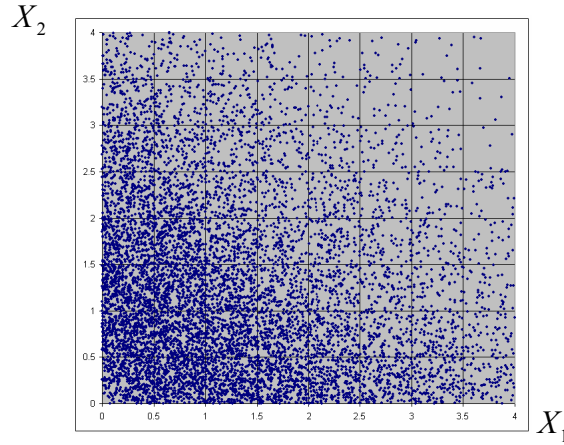
Institut für Mathematik, Schwerpunkt Versicherungs- und Finanzmathematik, Carl von Ossietzky Universität Oldenburg, Oldenburg, Deutschland
E-Mail: dietmar.pfeifer@uni-oldenburg.de

$$h_2(x, y) = \frac{f_0(x)f_1(y) + f_1(x)f_0(y) + (n-2)f_0(x)f_0(y)}{n}, \quad x, y \geq 0$$

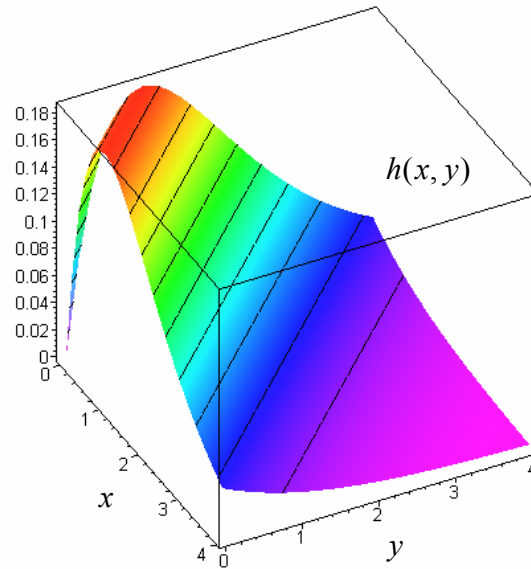
which implies that we have the following expressions for moments of random variables $\mathbf{X} = (X_1, \dots, X_n)$ with density h :

$$E(X_i) = \frac{n+1}{n}, \quad E(X_i^2) = \frac{2n+4}{n}, \quad Var(X_i) = \frac{n^2+2n-1}{n^2}, \quad Kov(X_i, X_j) = -\frac{1}{n^2}, \quad 1 \leq i < j \leq n$$

and correlation $\rho(X_i, X_j) = -\frac{1}{n^2+2n-1}$, $1 \leq i < j \leq n$. The following graph shows a scatterplot of 10,000 simulated pairs (X_1, X_2) for case $n = 2$:



The following graph shows the corresponding density for $n = 2$:



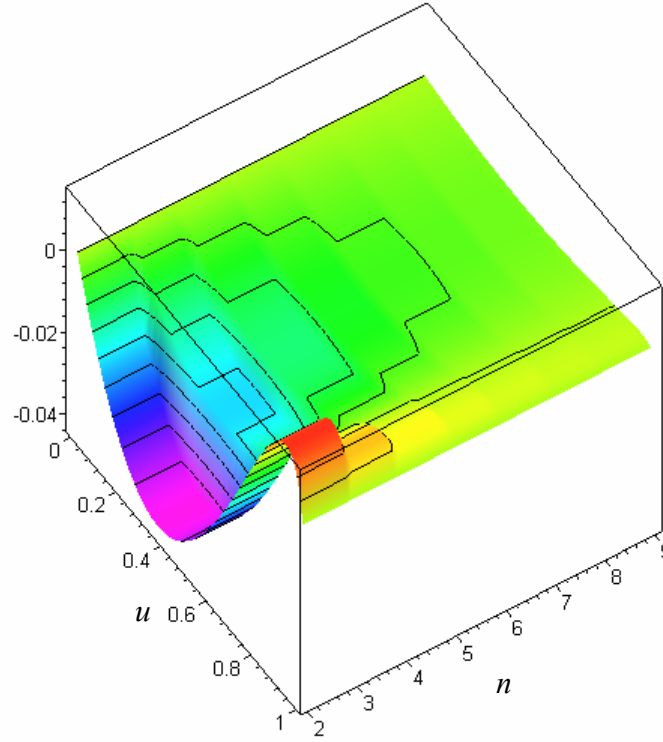
Unfortunately, the two-dimensional copulas C_2 derived from H_2 cannot be explicitly stated since the quantile function Q_1 pertaining to H_1 cannot be expressed in simple terms. But we

can approximate H_1 as follows: $H_1(x) = 1 - \left(1 + \frac{x}{n}\right) \exp(-x) \approx 1 - \exp\left(-\frac{n+1}{n}x\right)$, $x \geq 0$ from

which we get an approximation Q_1^* to the true quantile function Q_1 as

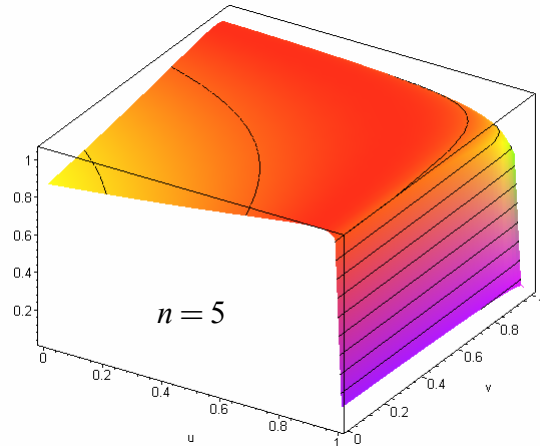
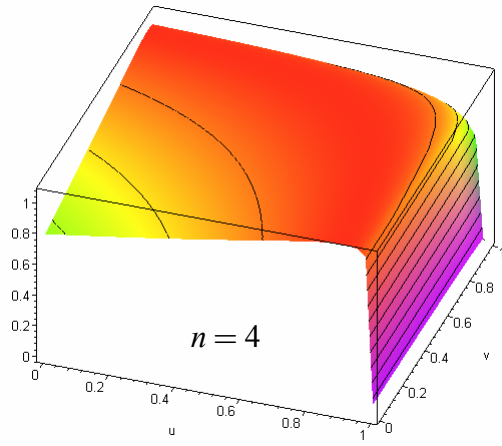
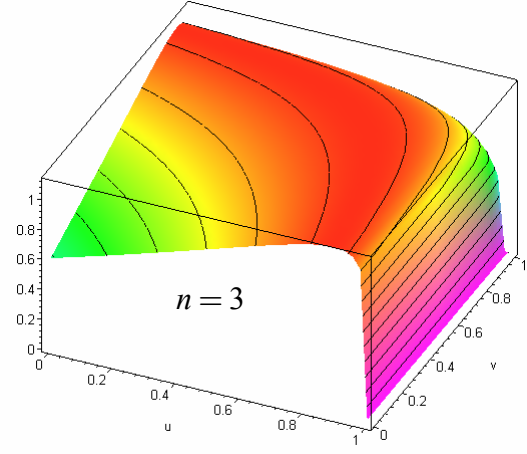
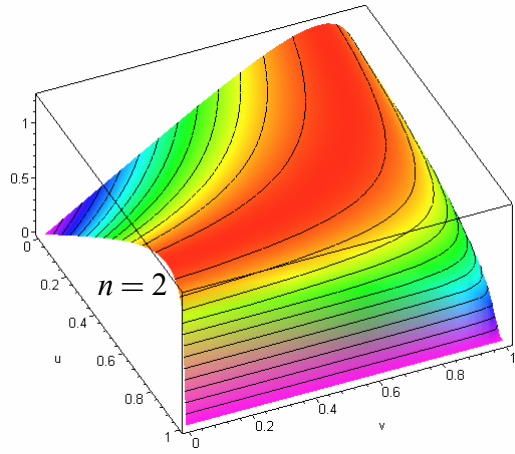
$Q_1^*(u, n) := -\frac{n}{n+1} \ln(1-u)$, $u > 0$, $n \geq 2$. The following 3d-graph shows the functions

$H_1(Q_1^*(u, n)) - u$ for $0 < u \leq 1$ and $2 \leq n \leq 9$:



As to be expected, the approximation error gets smaller for increasing values of n . We thus get the following approximation $C^*(u, v, n) := H_2(Q_1^*(u, n), Q_1^*(v, n))$ to the true copula $C(u, v) = H_2(Q_1(u, v), Q_1(u, v))$, $0 \leq u, v \leq 1$, according to Sklar's theorem (cf. [2], p. 47,

Corollary 2.10.10). The following graphs show the densities $c^*(u, v, n) = \frac{\partial^2}{\partial v \partial u} C^*(u, v, n)$ of $C^*(u, v, n)$ for various values of n :



Seemingly, with increasing n , $C^*(\bullet, \bullet, n)$ tends to the independence copula which is clear since $H_2(x, y) = \frac{F_0(x)F_1(y) + F_1(x)F_0(y) + (n-2)F_0(x)F_0(y)}{n}$ tends to $F_0(x)F_0(y)$ for increasing n , $x, y \geq 0$.

References

- [1] N.L. Johnson, S. Kotz, N. Balakrishnan (1994): Continuous Univariate Distributions, Volume 1, 2nd ed., Wiley, N.Y.
- [2] R.B. Nelsen (2006): An Introduction to Copulas. 2nd ed., Springer, N.Y.