

ANNUITY VALUATION BY COPULAS

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THE TWELFTH INTERNATIONAL CONFERENCE
ON FUZZY SET THEORY AND APPLICATIONS

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Core Syllabus for Actuarial Training in Europe

- basic probability theory
- random variables and related concepts
- correlation and regression analysis
- simulation methods

Actuaries strive

- to understand stochastic outcomes of financial security systems
- to estimate of joint life mortality and multidecrement models

Copula - very important concept in life insurance

Copula

- a copula is a function that connects univariate marginal distribution functions to their full multivariate distribution function
- copulas are useful for examining the dependence structure of multivariate random variables

The range of copulas applications

- civil engineering- reliability of analysis of highway bridges
- climate and weather related research
- analysis of extremas in financial assets and returns
- failure of paired organs in health science
- **human mortality in insurance (actuarial science)**
 - **mortalities of spouses**
 - **mortalities of parents and children**
 - **mortality of twins (identical or non-identical)**

Distribution functions which are useful for modeling of age at death

Gompertz distribution function, $G(m; \sigma)$

$$F(x) = 1 - e^{\left[e^{(-\frac{m}{\sigma})} \cdot (1 - e^{(\frac{x}{\sigma})}) \right]}$$

Weibull distribution function, $W(\gamma; c)$

$$F(x) = 1 - e^{-cx^\gamma}$$

Pareto distribution function, $Pa(\alpha; \lambda)$

$$F(x) = 1 - \left(\frac{\lambda}{\lambda + x} \right)^\alpha$$

Modeling of dependence

Uniform Gompertz distribution functions

$$u = F_1(x) = 1 - e \left[e^{(-\frac{m_1}{\sigma_1})} \cdot \left(1 - e^{(\frac{x}{\sigma_1})} \right) \right]$$

$$v = F_2(y) = 1 - e \left[e^{(-\frac{m_2}{\sigma_2})} \cdot \left(1 - e^{(\frac{y}{\sigma_2})} \right) \right]$$

estimation of individual parameters

$$\theta = (m_1, \sigma_1, m_2, \sigma_2)$$

Bivariate distribution function - Copula

$$H(x, y) = C(F_1(x), F_2(y))$$

basic property

It is clear that if $F_1(x)$, $F_2(y)$ and C are known, then H can be determined.

Sklar (1959) proved a converse:

"If H is known and $F_1(x)$, $F_2(y)$ are known and continuous, then C is uniquely determined".

Copula C belongs to the Archimedean class of copulas if

$$C_\phi(u, v, \alpha) = \phi^{-1}(\phi(u) + \phi(v))$$

a generator of the copula

$\phi :]0, 1] \rightarrow [0, \infty[$ is a convex, decreasing function satisfying $\phi(1) = 0$

estimation of individual parameters

$$\theta = (m_1, \sigma_1, m_2, \sigma_2, \alpha)$$

Characteristic of Archimedean copulas

Family of copulas	Parameter α	Generator $\phi(t)$	Kendall's τ (Spearman's ρ)
Gumbel	$\alpha > 1$	$(-\ln)^{\alpha}$	$\frac{\alpha-1}{\alpha}$ (no closed form)
Clayton	$\alpha > 0$	$\frac{t^{-\alpha}-1}{\alpha}$	$\frac{\alpha}{\alpha+2}$ (complicated form)
Frank	$\alpha \in \mathbb{R}$	$-\ln \left(\frac{e^{-\alpha t}-1}{e^{-\alpha}-1} \right)$	$1 - \frac{4}{\alpha}(1 - D_1(\alpha))$ $(1 - \frac{12}{\alpha}(D_2(-\alpha) - D_1(-\alpha)))$

Kendall's τ and Spearman's ρ - measures of the association between two variables (X, Y)

Kendall's τ

For each pair of observations (x_1, y_1) and (x_2, y_2)

we consider it concordant if $\frac{x_1 - x_2}{y_1 - y_2} > 0$ and discordant if $\frac{x_1 - x_2}{y_1 - y_2} < 0$

$$\tau = \frac{C - D}{\frac{1}{n}n \cdot (n - 1)}$$

Spearman's ρ

is the ordinary (Pearson) correlation coefficient of the transformed random variables $F_1(x)$ and $F_2(y)$.

Measures of the association between two variables expressed by copula function

Kendall's τ

$$\tau = 1 - 4 \int \int_{[0,1]^2} \frac{\partial C}{\partial u}(u, v) \frac{\partial C}{\partial v}(u, v) dudv$$

Spearman's ρ

$$\rho = 12 \int \int_{[0,1]^2} C(u, v) dudv - 3$$

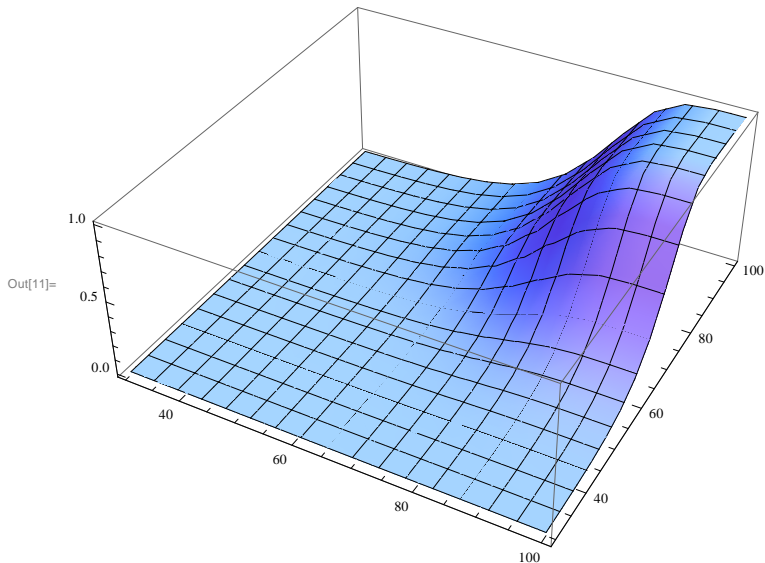
Frank's copula

$$C_F(u, v, \alpha) = \frac{1}{\alpha} \ln \left[1 + \frac{(e^{\alpha u} - 1) \cdot (e^{\alpha v} - 1)}{e^{\alpha} - 1} \right]$$

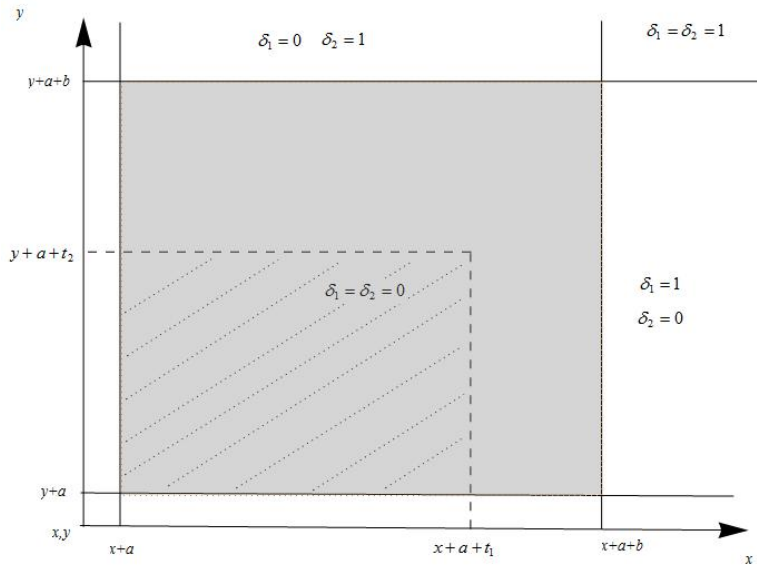
corresponding generator

$$\phi(t) = -\ln \frac{e^{-\alpha t} - 1}{e^{-\alpha} - 1}$$

Figure : Frank's copula



Maximum Likelihood Estimation



Conditional distribution function of lifetime random variables

$$H_T(t_1, t_2) = P(T_1 \leq t_1, T_2 \leq t_2 | T_1 > 0, T_2 > 0)$$

$$\frac{H(x + a + t_1, y + a + t_2) - H(x + a + t_1, y + a) - H(x + a, y + a + t_2) + H(x + a, y + a)}{1 - H(x + a, \infty) - H(\infty, y + a) + H(x + a, y + a)}$$

Maximum Likelihood Estimation

$$\begin{aligned} \ln L(x, y, t_1, t_2, \delta_1, \delta_2, a, b) &= \\ &= (1 - \delta_1) \cdot (1 - \delta_2) \cdot \ln h(x + a + t_1, y + a + t_2) + \\ &+ (1 - \delta_1) \cdot \delta_2 \cdot \ln (H_1(x + a + t_1, \infty) - H_1(x + a + t_1, y + a + b)) + \\ &+ \delta_1 \cdot (1 - \delta_2) \cdot \ln (H_2(\infty, y + a + t_2) - H_2(x + a + b, y + a + t_2)) + \\ &+ \delta_1 \cdot \delta_2 \cdot \\ &\ln (1 - H(x + a + b, \infty) - H(\infty, y + a + b) + H(x + a + b, y + a + b)) - \\ &- \ln (1 - H(x + a, \infty) - H(\infty, y + a) + H(x + a, y + a)) \end{aligned}$$

log-likelihood function for the data set

$$\ln \mathcal{L} = \sum_{i=1}^n \ln L(x_i, y_i, t_{1i}, t_{2i}, \delta_{1i}, \delta_{2i}, a_i, b_i)$$

Net single premium for independent lives aged x and y

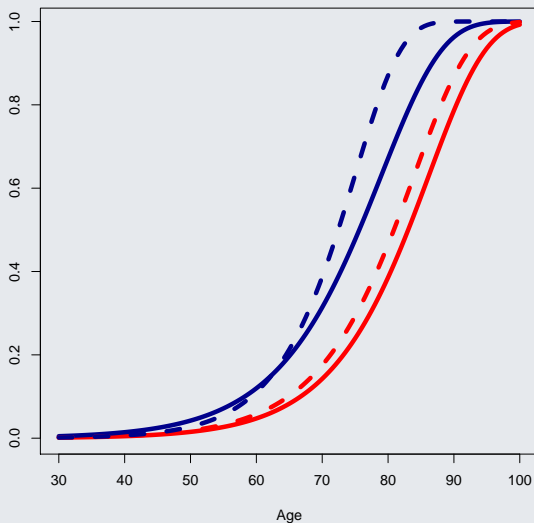
$$\ddot{a}_x = \sum_{t=0}^{\infty} v^t \cdot {}_t p_x \quad {}_t p_x = 1 - H_T(t, \infty)$$

$$\ddot{a}_y = \sum_{t=0}^{\infty} v^t \cdot {}_t p_y \quad {}_t p_y = 1 - H_T(\infty, t)$$

Net single premium for a joint and last-survivor annuity issued to lives aged x and y

$$\ddot{a}_{\bar{x}\bar{y}} = \sum_{t=0}^{\infty} v^t \cdot {}_t p_{\bar{x}\bar{y}} \quad {}_t p_{\bar{x}\bar{y}} = 1 - H_T(t, t), \quad a = 0$$

Figure : Distribution functions



Estimated parameters of the copula function

$$\theta = (m_1 = 75, \sigma_1 = 7, 0; m_2 = 84, \sigma_1 = 8, 5; \alpha = -0.13)$$

Yearly pension annuity from accumulated sum S

$$P(\alpha) = \frac{S}{\ddot{a}_{\bar{x}\bar{y}}(\alpha)} \quad P = \frac{S}{\ddot{a}_{\bar{x}\bar{y}}}$$

Ratios of Dependent to Independent Joint and last survivor pension annuities

Source: E. W. Frees, J. Carriere, E. Valdez: *Annuity valuation with dependent mortality*, 1995 (5 % p.a.)

	Female age		
Male age	60	65	70
60	0.95	0.94	0.94
65	0.96	0.94	0.93
70	0.97	0.95	0.94

Source: own construction (2 % p.a.)

	Female age		
Male age	60	65	70
60	0.76	0.79	0.84
65	0.76	0.88	0.83
70	0.87	0.85	0.85