Spatial Phenomena Exhibiting Extremes

Modelling Extremes with the Spatial Vine Copula

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Definition of the Spatial Vine Copula

Spatial dependence between locations may change with distance in *strength*:



The four bivariate spatial copulas used in the 5-dim. spatial vine copula. The copula families are denoted by: Gaussian "N", student "t", Clayton "C", Frank "F", Gumbel "G", Joe "J", survival Clayton "sC", survival Gumbel "sG", cubic-quadratic Sec. "CQ", product "I"). and shape between the CDFs of point pairs:

Usage of the Spatial Vine Copula

Predictions are obtained from the full distribution by means of any *p*-quantile (i.e. the *median*) or the *expected value*

$$\hat{x}_{m} = F^{-1}(C^{-1}(p|u_{1}, \ldots, u_{d})) \qquad \hat{x}_{m} =$$

 $= \int_{[0,1]} F^{-1}(u) \cdot c(u|u_1,\ldots,u_d) \mathrm{d} u.$

Simulations can be drawn from the conditional density at each location following a random path.

Uncertainties may follow any distribution and are given through a full but family free conditional distribution.

Application of the Spatial Vine Copula

Use case: Interpolation of the "Joker" dataset from the Spatial Interpolation Comparison 2004: simulated radiation including an accidental release of radioactive material. Interpolation results for the median spatial vine copula predictor:



Structural changes of the strength of dependence with distance. Asymmetric dependence structures (i.e. non-Gaussian) may be present. Spatial copulas can represent dependence structures that change with distance. Their density is given as convex combination of bivariate copulas:

$$c_h(u,v) := egin{cases} c_{1,h}(u,v) & , 0 \leq h < l_1 \ (1-\lambda_2)c_{1,h}(u,v) + \lambda_2 c_{2,h}(u,v) & , l_1 \leq h < l_2 \ & & : \ (1-\lambda_k)c_{k-1,h}(u,v) + \lambda_k \cdot 1 & , l_{k-1} \leq h < l_k \ 1 & , l_k \leq h \end{cases}$$

where $\lambda_j := \frac{h - l_{j-1}}{l_j - l_{j-1}}$, *h* denotes the separating distance and l_1, \ldots, l_k denote the representative distances of the bins (e.g. mean distance of all involved point pairs).



Validation: 808 additionally simulated locations are used to validate the interpolation based on 200 locations.

approach	MAE	RMSE	ME	COR
spatial vine copula	14.5	67.6	-6.1	0.60
TG log-kriging	20.8	78.2	-2.1	0.39
residual kriging	21.1	75.6	5.2	0.43

Conditional CDF describe the uncertainties of the prediction:



Spatial vine copulas join pair-wise *spatial copulas* into a multivariate distribution of a local neighbourhood through a *vine*:



Graphical representation of a spatial vine copula. Each tree *i* has its own bivariate spatial copula $c_{h_{i-1}}$ describing the changing dependence between pairs of locations.

Its density is given through the product of all involved bivariate copula densities:

$$c_{\mathbf{h}}(u_0,\ldots,u_d) = \prod_{i=1}^d c_{h_0(i)}(u_0,u_i) \cdot \prod_{j=1}^{d-1} \prod_{i=1}^{d-j} c_{h_j(j+i)}(u_{j|0,\ldots,j-1},u_{j+i|0,\ldots,j-1})$$

where $u_i = F_i(Z(s_i))$ for $0 \le i \le d$ and

$$u_{j+i|0,...,j-1} = F_{h_{j-1}(j+i)}(u_{j+i}|u_0,...,u_{j-1})$$

= $\frac{\partial C_{h_{j-1}(j+i)}(u_{j-1|0,...,j-2},u_{j+i|0,...,j-2})}{\partial u_{j-1}}$

Conditional CDFs of the median spatial vine copula predictor and residual kriging at an extreme (left) and a background (right) location.

Software and code is available as R-package *spcopula* on r-forge (talk *S2.2* on Thursday).

Conclusions

The spatial vine copula ...

- flexibly describes spatial dependence of local neighbourhoods.
- is able to *capture extremes*.
- allows to use *any marginal distribution*.
- *outperforms* "classical" geostatistical approaches (in terms of MAE even the best SIC2004 participant).

• directly provides *conditional distributions describing uncertainties*.

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