



Bounding the dependence measures for spatial uncertainties

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Introduction

Dependence
measures

Spatial dependence

Academic example

Conclusions

- 1 Introduction
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Representation of spatial uncertainty in numerical modelling

→ **Needed at *input* side of an analysis:**

E.g., material properties, loads

Framework to capture the available information?

- (Spatial variation of) Bounds on parameter
- Dependency between local values of parameter: limit the possible realisations in between the bounds.

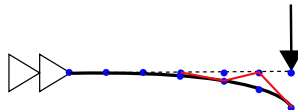
→ **Needed at *output* side of an analysis:**

E.g., displacement field, temperature field

How to accurately represent deterministic realisation?

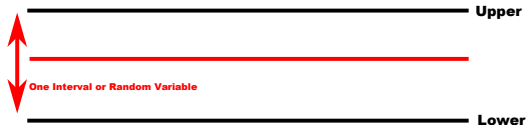
How to represent uncertainty without conservatism?

- Predicting correct bounds on local values
- Avoid impossible realisations



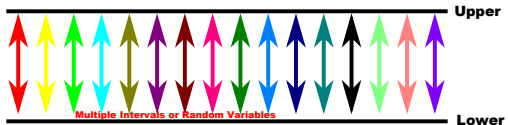
Representation of spatial uncertainty in numerical modelling

- Assuming uniformity or homogeneity, total dependence:



This representation is **not sufficient**.

- Assuming total independence:



This representation is **not necessary** and **infeasible**.

Representation of spatial uncertainty in numerical modelling

What is needed?

- **Dependence measure**
- **Realisations**
- **Uncertainty propagation**
 - Types:
 - scalar \rightarrow field
 - field \rightarrow scalar
 - field \rightarrow field
 - Optimisation problem: what to optimise?
- **Representation of output**
 - Bounds with dependence measure \rightarrow realisations
 - Analytical formulation

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Non-probabilistic
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Dependence measures

“We must be clearer about the abused word *dependence* and its relatives.” [Drouet Mari and Kotz, 2001]

- ‘Perfect’ dependence: $x = a \rightarrow y = b$
- ‘Flexible’ dependence: $x \in x^s \subset x^S \rightarrow y \in y^s \subset y^S$

“... an essential part of uncertainty analysis is the analysis of dependence.” [Kurowicka and Cooke, 2006]

“... , the only means to establish a relationship between variables was to deduce a causative connection. There was no way to discuss - let alone measure - the association between variables that lack a cause-effect relationship.” [Rodgers and Nicewander, 1988]



Dependence measures

Probabilistic dependence measures

Probabilistic dependence measures:

- Scalar aggregate (global) measure:
 - Product moment correlation
 - Rank correlation
 - Kendall's tau
 - Relative entropy
- Thorough (local) measure:
 - Copula

Dependence measures

Probabilistic dependence measures

Product moment correlation

$$\rho(X, Y) \equiv \frac{E(XY) - E(X)E(Y)}{\sigma_X \sigma_Y}$$

- Depends on marginal distributions
- $-1 \leq \rho(X, Y) \leq 1$
- min & max (not necessarily -1 and 1) when countermonotonic or comonotonic
- Invariant under linear strictly increasing transformations
- X and Y are independent $\rightarrow \rho(X, Y) = 0$

Dependence measures

Probabilistic dependence measures

Rank correlation

$$\rho_r(X, Y) \equiv \rho(F_X(X), F_Y(Y))$$

- Independent of marginal distributions
- $-1 \leq \rho_r(X, Y) \leq 1$
- min & max when countermonotonic or comonotonic
- Invariant under non-linear strictly increasing transformations
- X and Y are independent $\rightarrow \rho_r(X, Y) = 0$

Elucidation

$$\pi_{\frac{1}{2}}(X, Y) \equiv P\left(F_X(X) > \frac{1}{2} | F_Y(Y) > \frac{1}{2}\right)$$

- $\pi_{\frac{1}{2}}(X, Y) = 0 \rightarrow \rho_r(X, Y) = -1$
- $\pi_{\frac{1}{2}}(X, Y) = 1/2 \rightarrow \rho_r(X, Y) = 0$
- $\pi_{\frac{1}{2}}(X, Y) = 1 \rightarrow \rho_r(X, Y) = 1$

Dependence measures

Probabilistic dependence measures

Kendall's tau

$$\tau(X, Y) \equiv P[(X_1 - X_2)(Y_1 - Y_2) \geq 0] - P[(X_1 - X_2)(Y_1 - Y_2) < 0]$$

- Independent of marginal distributions
- $-1 \leq \tau(X, Y) \leq 1$
- Invariant under non-linear strictly increasing transformations
- X and Y are independent $\rightarrow \tau(X, Y) = 0$

Dependence measures

Probabilistic dependence measures

Relative entropy

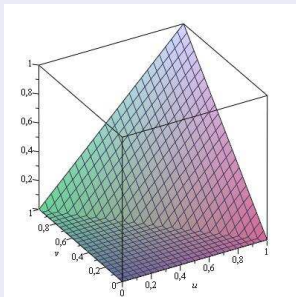
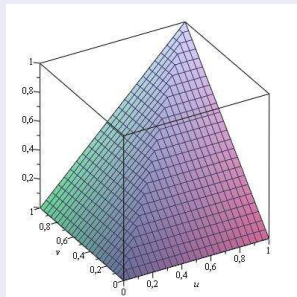
$$\delta_{X,Y} \equiv \int \int f(x,y) \log \left(\frac{f(x,y)}{f_1(x)f_2(y)} \right) dx dy$$

- Entropy of $f(x,y)$ is compared with maximum attainable entropy when X and Y are independent
- X and Y are independent $\rightarrow \delta_{X,Y} = 0$
- X and Y are perfect dependent $\rightarrow \delta_{X,Y}$ tends to ∞

Copula

$$F_{XY}(x, y) = C(F_X(x), F_Y(y))$$

- Separates effect of dependence and marginal distributions
- Cumulative distribution function with uniform marginals on $[0, 1]$
- $C_L = \max(u + v - 1, 0) \leq C(u, v) \leq C_U = \min(u, v)$ with $(u, v) \in [0, 1]^2$

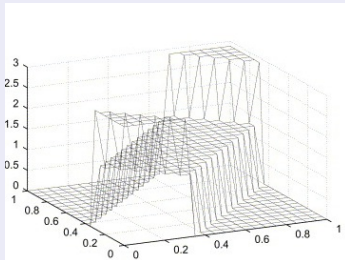
 C_L  C_U

Dependence measures

Probabilistic dependence measures

Most promising two copulae

- Diagonal band copula



- Minimum information copula: unique bivariate joint distribution with

- marginal distribution uniform on $I = \left[-\frac{1}{2}, \frac{1}{2}\right]$
- rank correlation $\rho_r \in [-1, 1]$
- minimal information relative to uniform distribution among all distributions with rank correlation ρ_r
- no closed functional form

Dependence measures

Non-probabilistic dependence measures

Non-probabilistic dependence measures:

- Independence and noninteraction dominate

Hartley measure of uncertainty

$$H(r_E) = \log_2 |E|, \quad r_E(x) = \begin{cases} 1 & \text{when } x \in E \\ 0 & \text{when } x \notin E \end{cases}$$

Information transmission

$$T_H(X, Y) = H(X) + H(Y) - H(X \times Y)$$

- Noninteractive $\rightarrow T_H(X, Y) = 0$
- $H(X|Y) = H(Y|X) = 0 \rightarrow$ maximum of $T_H(X, Y)$

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Spatial dependence

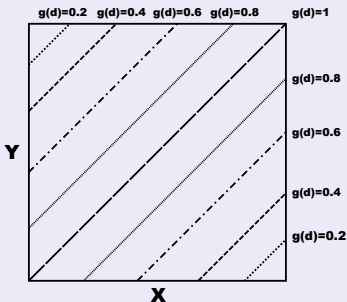
[Link to spatial dependence](#)

Link to spatial dependence:

- A dependence description as a function of the distance between points
 - Prob: Covariance structure of Random Field
 - Well developed discretisation methods
 - Criteria to assess the discretisation
 - Non-prob: Domain of influence

Domain of influence

- Inspired by diagonal band copula



- d is distance between two points
- $g(d) \rightarrow$ set $R \subseteq X \times Y$ of possible values
- Other shapes for R possible

Domain of influence: $g(d)$

- $g(d) : d \rightarrow [-\infty, 1]$
- $g(d) \leq 0$ then $R = X \times Y$
- $g(d) = 1$ then R reduces to the single line $X = Y$.
- Conditions on $g(d)$:
$$\begin{cases} g(0) = 1 \\ g(d_1) > g(d_2) \quad \text{for } d_1 < d_2 \end{cases}$$
- Example: $g(d) = 1 - \frac{d}{a}$
 $a > 0$, a parameter specifying the domain of influence.
 - $d < a$ then $R \subset X \times Y$
 - $d > a$ then $R = X \times Y$

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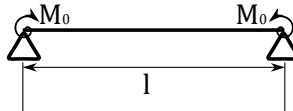
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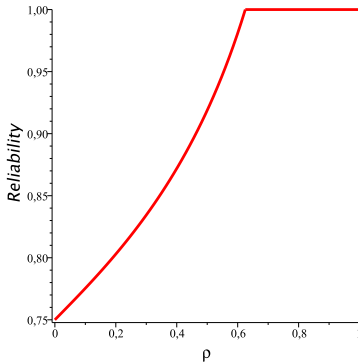
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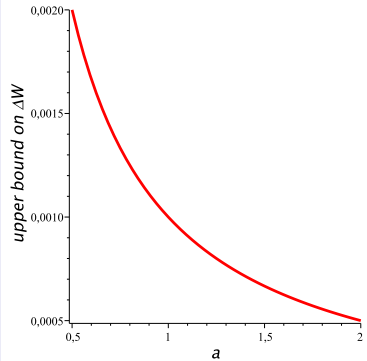
Academic example:

- Input:
 - uncertain flexibility, 2 point discretisation
 - interval on correlation length
 - interval on domain of influence a
- Output: displacement difference $\Delta W(x_1, x_2)$
 - Prob: reliability
 - Non-prob: upper bound

Reliability



Upperbound



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What was given:

- Overview of dependence measures
- Link between set definition and spatial uncertainty

What is still needed:

- Non-probabilistic dependence measures needed
- Dependence measure \rightarrow realisations:
 - criteria for realisations



International Conference on Noise and Vibration Engineering

The PMA Noise & Vibration research group organises :



The biennial ISMA conference on Noise and Vibration Engineering. [ISMA2012](#) will be organised on **September 17-19, 2012** in Leuven (Belgium), in conjunction with USD2012.



The 4th edition of the International Conference on Uncertainty in Structural Dynamics. [USD2012](#) will be organised on **September 17-19, 2012** in Leuven (Belgium), in conjunction with ISMA2012.



The annual ISMA course : "Modal Analysis, Theory and Practice". [ISMA37](#) will be organised on **September 20-21, 2012** in Leuven (Belgium).



The annual ISAAC course : "Advanced Techniques in Applied and Numerical Acoustics". [ISAAC23](#) will be organised on **September 20-21, 2012** in Leuven (Belgium).



The course : "Verification & Validation of Structural Dynamics Models". [V&VSDM](#) will be organised on **September 20-21, 2012** in Leuven (Belgium) and is taught by dr. F. Hemez of Los Alamos Dynamics, L.L.C.

The latest information

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