

Multivariate Models and Confidence Intervals: A Local Random Set Approach

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Research Team (imprecise probabilities):

- Michael Oberguggenberger (head of unit)
- Bernhard Schmelzer (next presentation)
- Thomas Fetz

My research area:

- Propagating uncertainty through a mapping.
- Imprecise probabilities and independence.
- Starting: [Bayesian Networks](#) linked with [Geographic Information Systems](#) (GIS) in collaboration with civil engineers of our faculty.



Families of Confidence Intervals / The Univariate Case

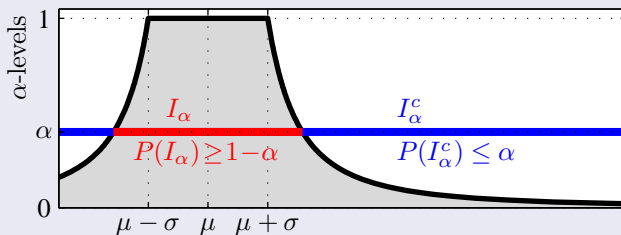
Starting point: Non-parametric models, Tchebycheff

Given: Variable X with $\mu = E(X)$ and $\sigma^2 = V(X)$ as sole information.

Generating a nested family $\mathbf{I} = \{I_\alpha\}_{\alpha \in (0,1]}$ of **confidence intervals**

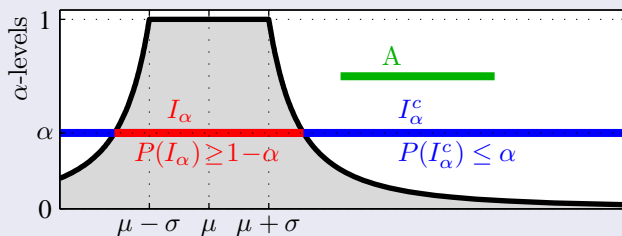
$$I_\alpha = \left[\mu - \frac{\sigma}{\sqrt{\alpha}}, \mu + \frac{\sigma}{\sqrt{\alpha}} \right], \quad \alpha \in (0, 1]$$

using **Tchebycheff's inequality** $P(|X - \mu| > \frac{\sigma}{\sqrt{\alpha}}) \leq \alpha$.



Questions

- What is it? Is it a random set or a fuzzy set?
- What is the upper probability $\bar{P}(A)$ of an event A ?
What is its interpretation with respect to confidence intervals?
- What happens in the multivariate case?

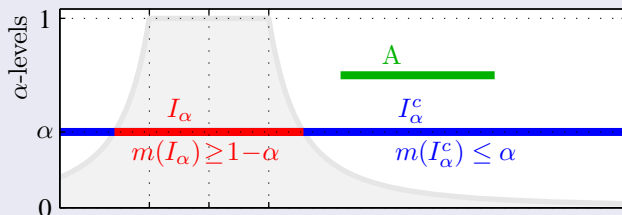


Looking at a single confidence interval

Equipping the two intervals I_α and I_α^c with weights

$$m(I_\alpha) = P(I_\alpha) \quad \text{and} \quad m(I_\alpha^c) = P(I_\alpha^c)$$

we get a **local random set at level α** .



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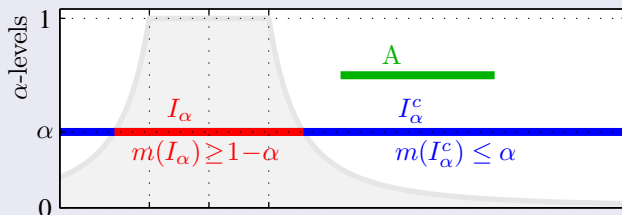
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The **local upper probability $\bar{P}_\alpha(A)$ at level α** for an event A is

$$\bar{P}_\alpha(A) = m(I_\alpha) \chi(A \cap I_\alpha \neq \emptyset) + m(I_\alpha^c) \chi(A \cap I_\alpha^c \neq \emptyset).$$

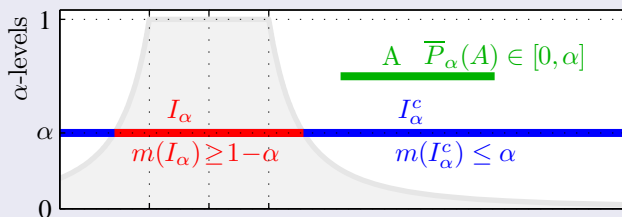
(χ indicator function)



Three different cases for an event A

(i) $A \cap I_\alpha = \emptyset$	$\bar{P}_\alpha(A) \in [0, \alpha]$
(ii) $A \cap I_\alpha^c = \emptyset$	$\bar{P}_\alpha(A) \in [1 - \alpha, 1]$
(iii) $A \cap I_\alpha \neq \emptyset$ and $A \cap I_\alpha^c \neq \emptyset$	$\bar{P}_\alpha(A) = 1$

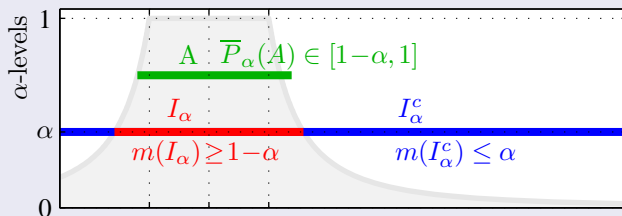
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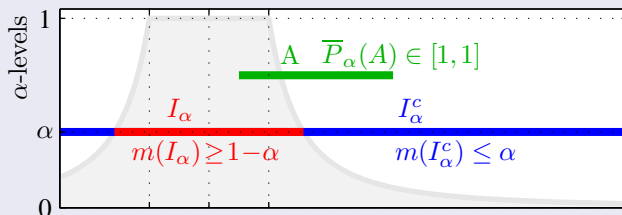
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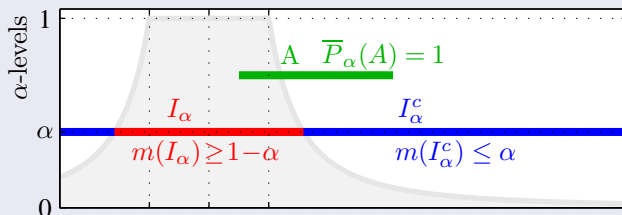
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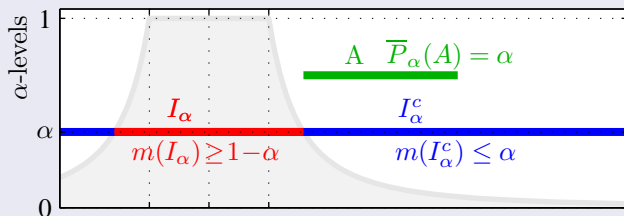
To avoid interval-valued $\bar{P}_\alpha(A)$ we take always the **upper bounds**.



Most interesting case (i)

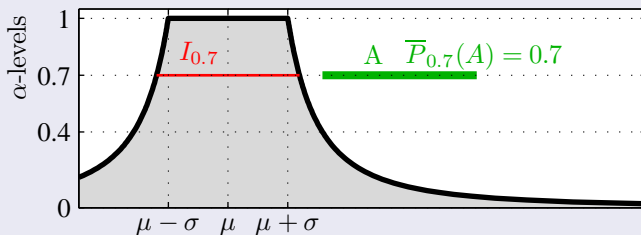
- A has the role of the “bad and undesired” event.
- Meaning:

If A is outside the confidence interval I_α at confidence level $1 - \alpha$, then we can say for sure that A occurs only with probability α , at most.



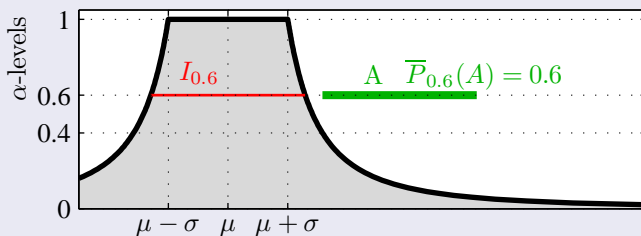
Formula for the upper probability $\bar{P}(A)$

$$\bar{P}(A) = \inf_{\alpha \in (0,1]} \bar{P}_\alpha(A) = \inf_{\alpha \in (0,1]} \{\alpha : I_\alpha \cap A = \emptyset\}$$



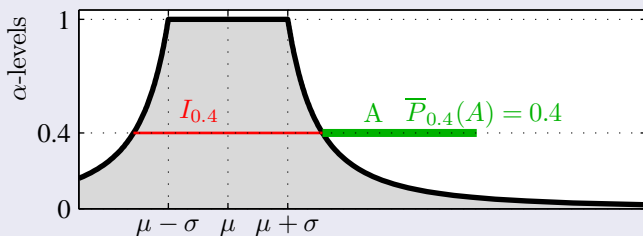
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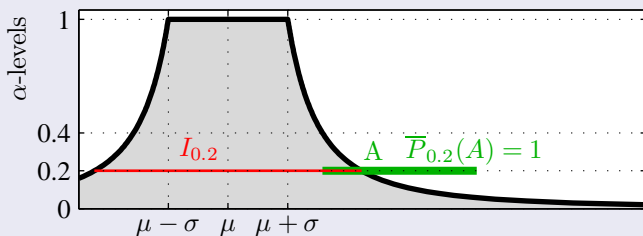
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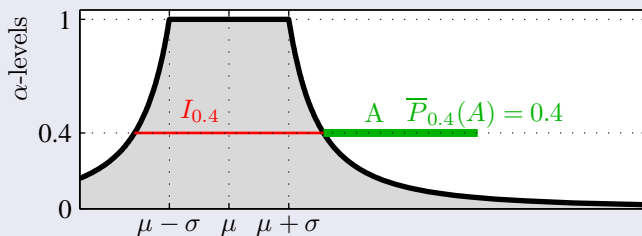


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Interpretation of I as random set or fuzzy set

$$\bar{P}(A) = \text{Plausibility}(A) = \text{Possibility}(A)$$



Goal

A **formula** for the upper probability for given families $\mathbf{I}_1, \dots, \mathbf{I}_n$ of confidence intervals **similar to the univariate version**.

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Short preview

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multivariate

$$\bar{P}_\ell^S(A) = \inf_{\alpha \in S} \{ \ell(\alpha) : J_\alpha \cap A = \emptyset \}$$

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Possibilities of choice

- 1 For the **set of confidence intervals** considered to be **combined**.
- 2 For the **weights** used for the local joint random set.

Combination of marginal confidence intervals

$\mathbf{J} = \{J_{\alpha}\}_{\alpha \in S}$ is the family of all **joint confidence sets**

$$J_{\alpha} = I_{1,\alpha_1} \times \cdots \times I_{n,\alpha_n}, \quad I_{k,\alpha_k} \in \mathbf{I}_k$$

with $\alpha = (\alpha_1, \dots, \alpha_n)$ depending on the set S of indices α :

- 1 Random set independence like: $S = S_R = (0, 1]^n$.
- 2 Fuzzy set independence like: $S = S_F = \{\alpha \in (0, 1]^n : \alpha_1 = \cdots = \alpha_n\}$

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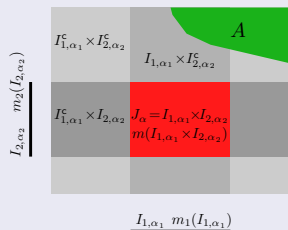
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Local upper probability



Event A with $J_{\alpha} \cap A = \emptyset$.

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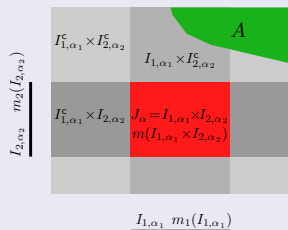
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$$\begin{aligned} \bar{P}_\alpha(A) &\leq \bar{P}((I_{1,\alpha_1} \times \cdots \times I_{n,\alpha_n})^c) = \\ &= 1 - m(I_{1,\alpha_1} \times \cdots \times I_{n,\alpha_n}). \end{aligned}$$

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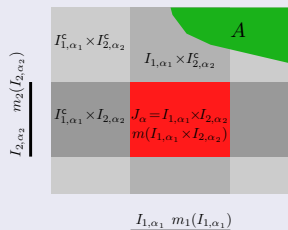
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Worst case.

Random set independence

Product of the marginal weights: $m(I_{1,\alpha_1} \times \dots \times I_{n,\alpha_n}) = \prod_{i=1}^n m_i(I_{i,\alpha_i})$.

Leads to

$$\bar{P}_\alpha(A) = 1 - \prod_{i=1}^n (1 - \alpha_i).$$

- Used if the uncertain variables are independent.

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Lower / upper bounds

Lower and upper bounds of Fréchet for the joint weights:

$$\max \left(\sum_{i=1}^n m(I_{i,\alpha_i}) - n + 1, 0 \right) \leq m(I_{1,\alpha_1} \times \dots \times I_{n,\alpha_n}) \leq \min_{i=1,\dots,n} m(I_{i,\alpha_i}).$$

Leads to

$$\max_{i=1,\dots,n} (\alpha_i) \leq \bar{P}_\alpha(A) \leq \min(\alpha_1 + \dots + \alpha_n, 1).$$

- Used if nothing is known about interactions between the variables.

Level function $\ell(\alpha)$

The different approaches have **only an influence on the level**

$$\ell(\alpha) = \begin{cases} \max_{i=1, \dots, n} (\alpha_i) & \text{lower bound,} \\ 1 - \prod_{i=1}^n (1 - \alpha_i) & \text{random set independence,} \\ \min(\alpha_1 + \dots + \alpha_n, 1) & \text{upper bound} \end{cases}$$

of the joint confidence set J_α , but not on the confidence set itself.

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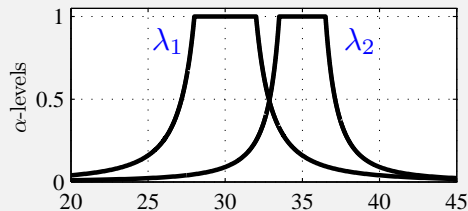
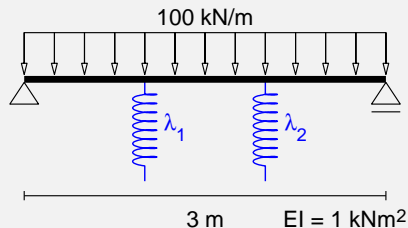
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Formula for the upper probability similar to the univariate case

$$\bar{P}_\ell^S(A) = \inf_{\alpha \in S} \{ \ell(\alpha) : J_\alpha \cap A = \emptyset \}.$$

Numerical Example

Beam bedded on two springs with uncertain spring constants



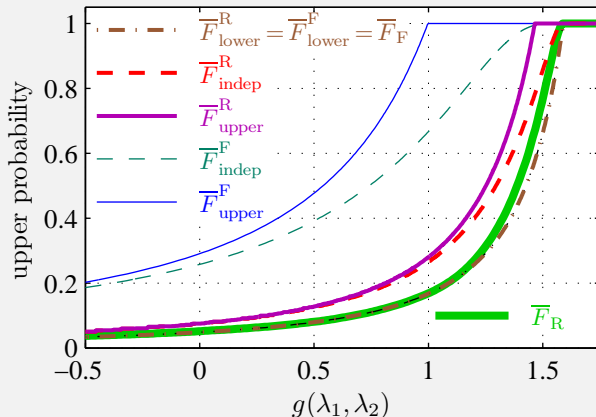
Criterion of failure of beam: $g(\lambda_1, \lambda_2) \leq 0$

Failure function

$$g(\lambda_1, \lambda_2) = M_{\text{yield}} - \max_{x \in [0, 3]} |M(x, \lambda_1, \lambda_2)|$$

- $M(x, \lambda_1, \lambda_2)$ is the bending moment at x depending on λ_1, λ_2 .
- $M_{\text{yield}} = 12 \text{ kNm}$ is the elastic limit moment.

Upper probability distributions $\overline{F}_\ell^S(g(\lambda_1, \lambda_2))$



The **upper probabilities of failure** are the results at $g(\lambda_1, \lambda_2) = 0$.

Ordering of the upper probabilities

$$\bar{P}_F(A) = \bar{P}_{\text{lower}}^R(A) = \bar{P}_{\text{lower}}^F(A)$$

$$\bar{P}_R(A) \leq \bar{P}_{\text{indep}}^R(A) \leq \bar{P}_{\text{indep}}^F(A)$$

$$\bar{P}_U(A) \leq \bar{P}_{\text{upper}}^R(A) \leq \bar{P}_{\text{upper}}^F(A)$$



Classical approaches

Notation	
\bar{P}_F	fuzzy set independence
\bar{P}_R	random set independence
\bar{P}_U	unknown interaction / Fréchet

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All possible combinations of confidence intervals, $S = S_R$

Notation	level $\ell(\alpha)$	
$\overline{P}_{\text{lower}}^R$	$\max_{i=1, \dots, n} (\alpha_i)$	lower Fréchet bound
$\overline{P}_{\text{indep}}^R$	$1 - \prod_{i=1}^n (1 - \alpha_i)$	random set independence
$\overline{P}_{\text{upper}}^R$	$\min(\sum_{i=1}^n \alpha_i, 1)$	upper Fréchet bound

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Combinations of intervals of the same level α only, $S = S_F$

Notation	level $\ell(\alpha)$	
$\overline{P}_{\text{lower}}^F$	α	lower Fréchet bound
$\overline{P}_{\text{indep}}^F$	$1 - (1 - \alpha)^n$	random set independence
$\overline{P}_{\text{upper}}^F$	$\min(n\alpha, 1)$	upper Fréchet bound