

# Measuring Experts' performance in assessing dependence: experiences and open questions.

- Motivation
- Conditional probabilities of exceedance
- Ratios of rank correlations
- Determinant correlation matrix
- Dependence Calibration Score

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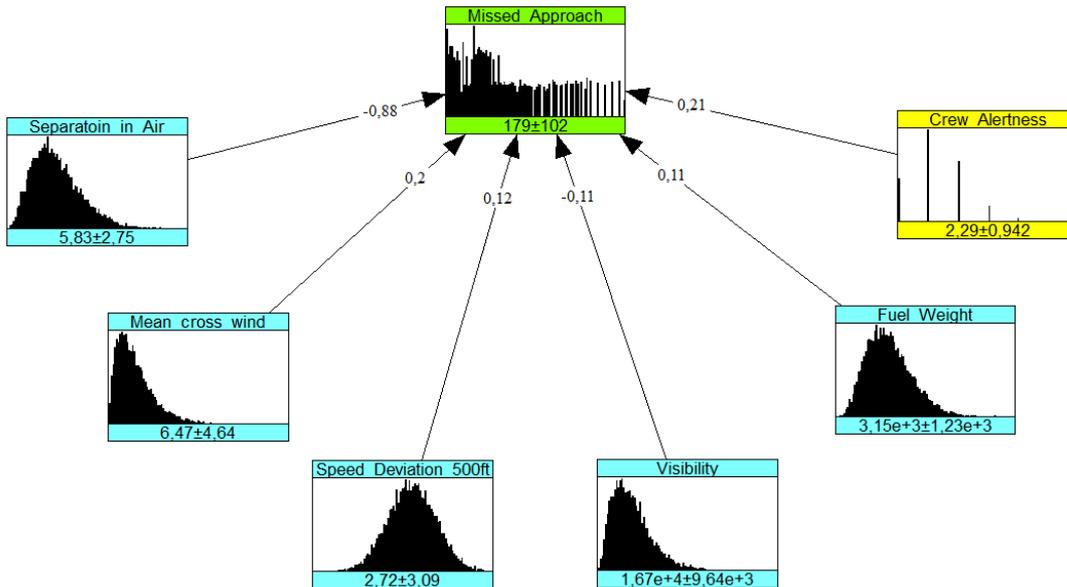
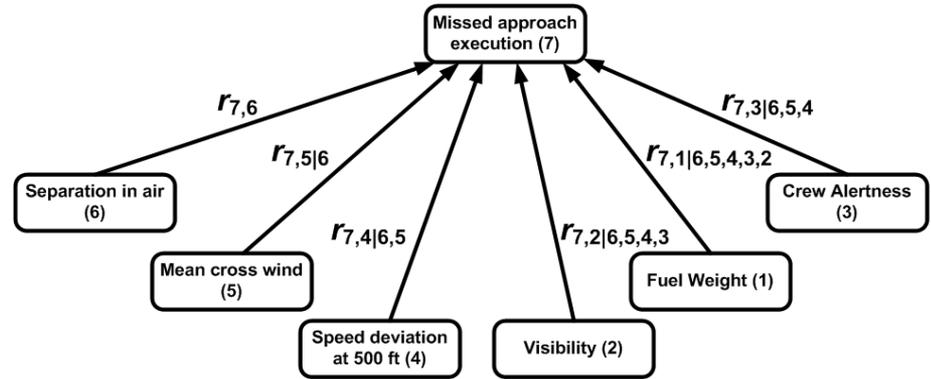
DEPENDENCE ELICITATION IS ESSENTIALLY DIFFERENT THAN UNCERTAINTY ELICITATION.

USUAL MEASURES FOR UNCERTAINTY ELICITATION ARE NOT SUFFICIENT.

# MOTIVATION

- Not always all JOINT data available.

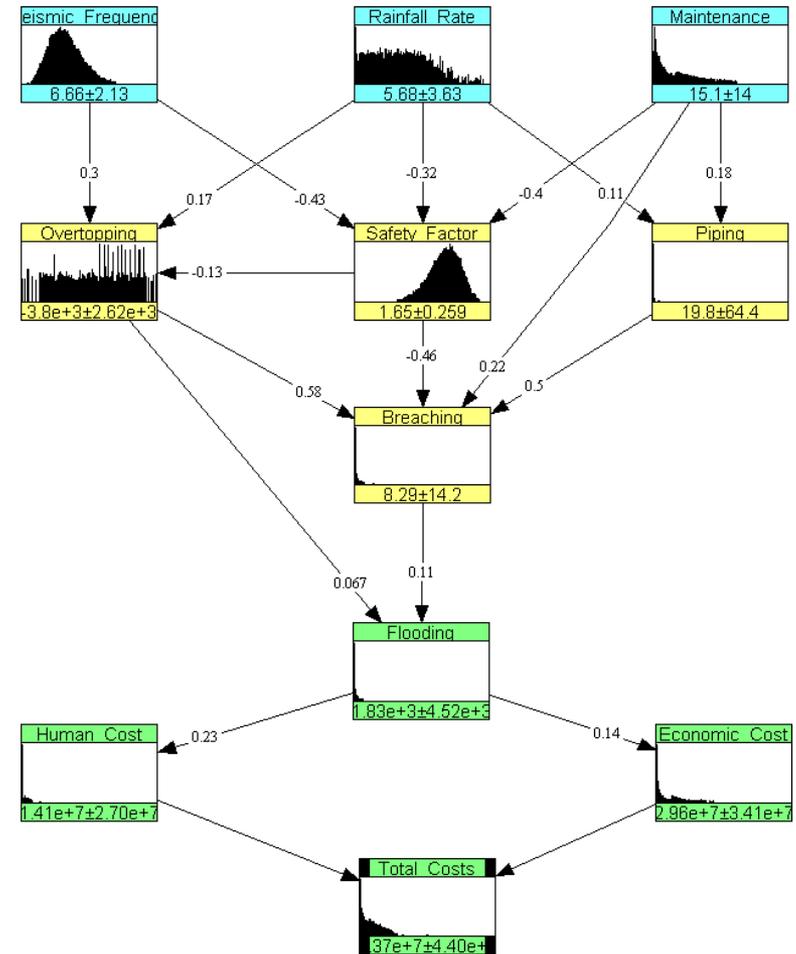
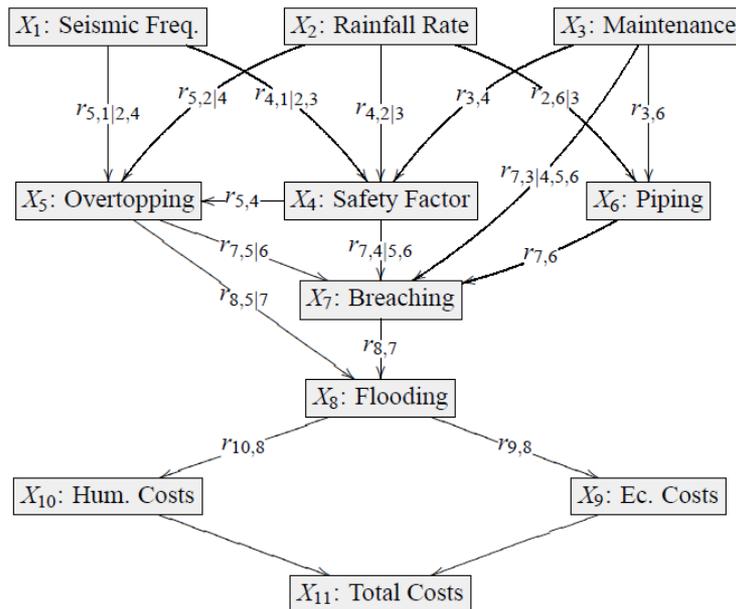
## CONTROLLED FLIGHT INTO TERRAIN



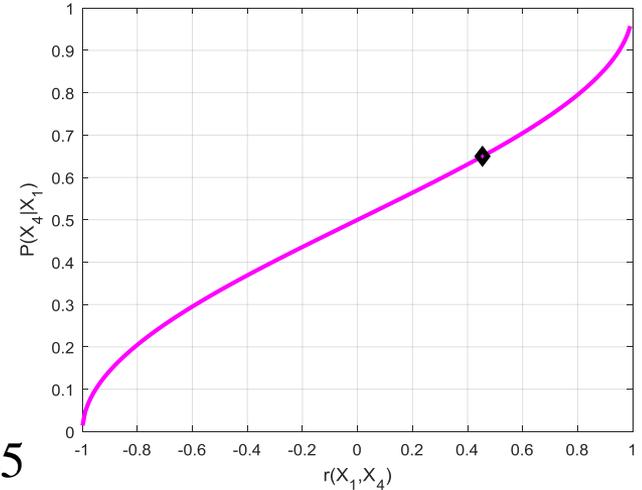
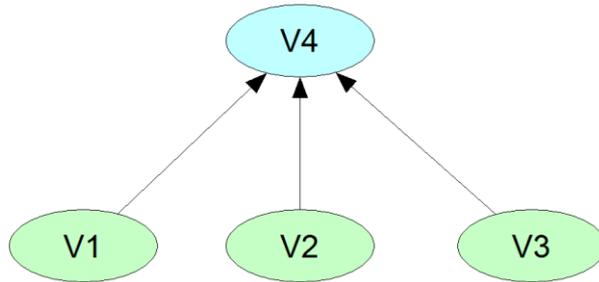
- ›  $P(7 > \text{med} | 6 > \text{med})$
- ›  $P(7 > \text{med} | 6 > \text{med}, 5 > \text{med}) \dots$
- ›  $P(7 > \text{med} | 1 > \text{med}, 2 > \text{med}, 3 > \text{med}, 4 > \text{med}, 5 > \text{med}, 6 > \text{med})$
- ›  $R(7,6)$
- ›  $R(7,5)/R(7,6) \dots$
- ›  $R(7,1)/R(7,6)$

# EARTH DAMS IN MEXICO

- ▶ 4 experts
- ▶ 16 (conditional) rank correlations
- ▶ Ratios of rank correlations

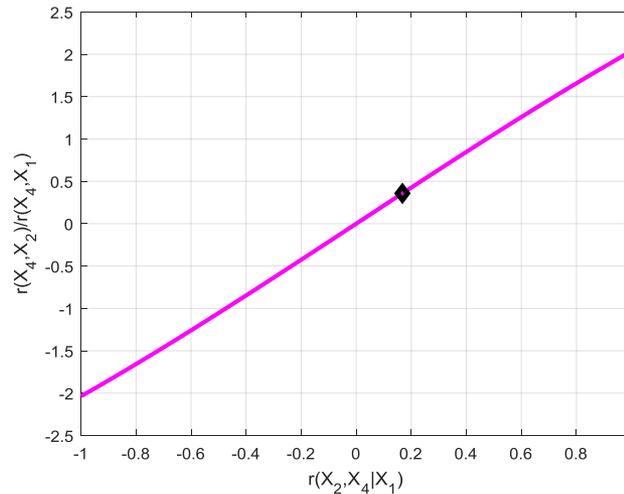
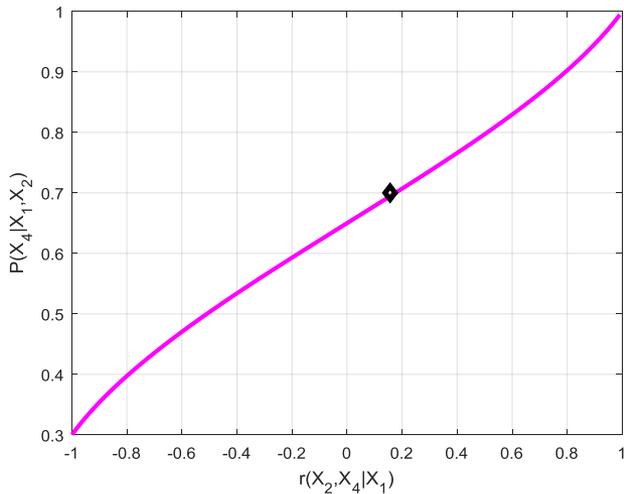


# BN DEPENDENCE QUANTIFICATION



$$P_1^{e_1} = 0.65 \rightarrow r_{4,1}^{e_1} = 0.45 \quad P_1^{e_2} = 0.65 \rightarrow r_{4,1}^{e_2} = 0.45$$

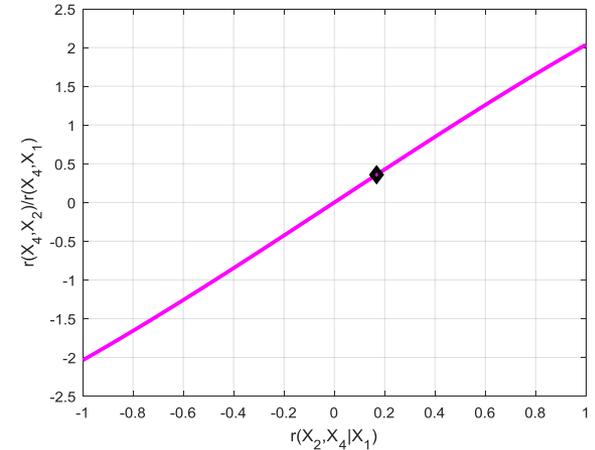
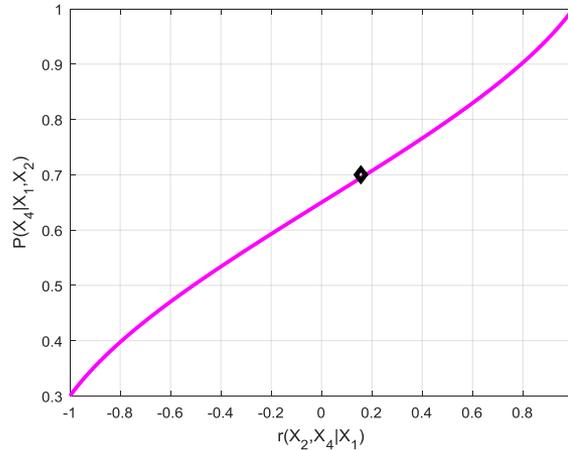
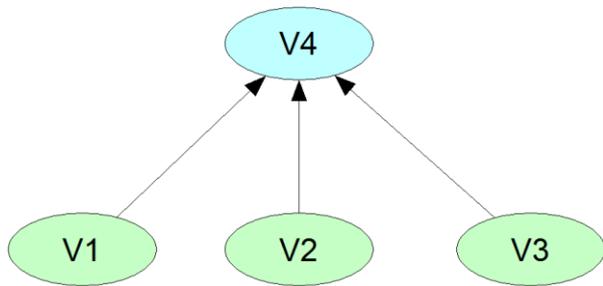
$$P_2^{e_1} = 0.70 \rightarrow r_{4,2|1}^{e_1} = 0.18 \quad R_2^{e_2} = 0.35 \rightarrow r_{4,2|1}^{e_2} = 0.18$$



	V1	V2	V3	V4
V1	1	0	0	0,45
V2	0	1	0	0,16
V3	0	0	1	0
V4	0	0	0	1

- › Cond. Prob. or
- › Rat. of Rank corr.
- › Are Bounded
- › Depends on the corr. Matrix elements
- › Positive definite

# BN DEPENDENCE QUANTIFICATION



$$P_1^{e_1} = 0.65 \rightarrow r_{4,1}^{e_1} \approx 0.44$$

$$P_1^{e_2} = 0.65 \rightarrow r_{4,1}^{e_2} \approx 0.44$$

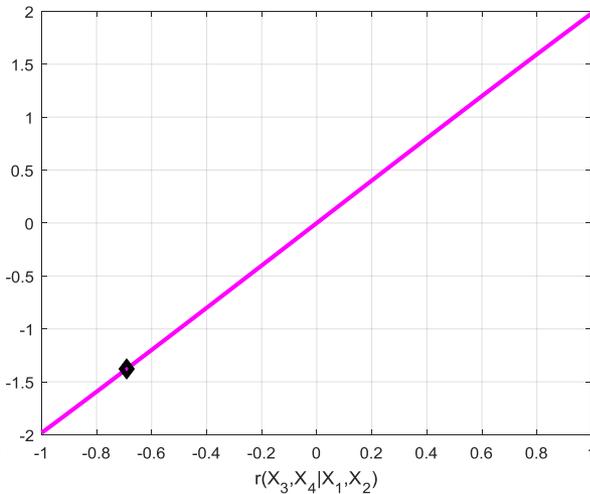
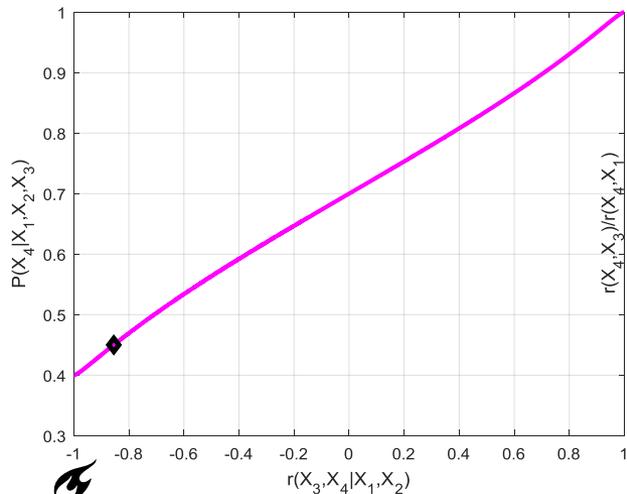
$$P_2^{e_1} = 0.70 \rightarrow r_{4,2|1}^{e_1} \approx 0.17$$

$$R_2^{e_2} = 0.36 \rightarrow r_{4,2|1}^{e_2} \approx 0.17$$

$$P_3^{e_1} = 0.50 \rightarrow r_{4,3|2,1}^{e_1} \approx -0.69$$

$$R_2^{e_2} = -1.38 \rightarrow r_{4,2|1}^{e_2} \approx -0.69$$

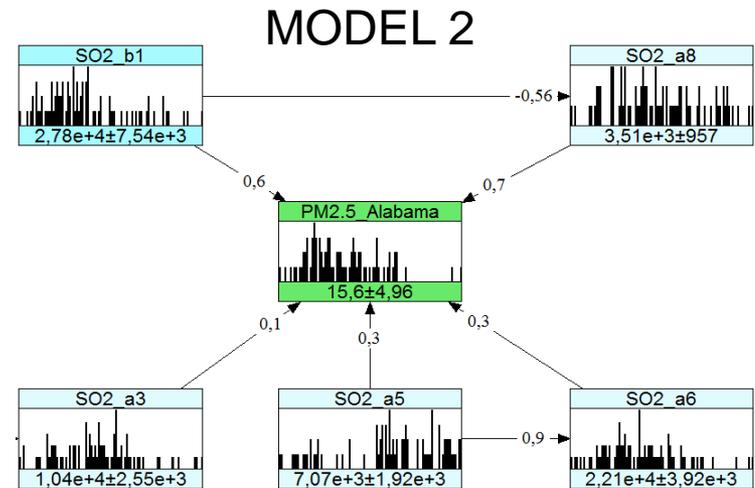
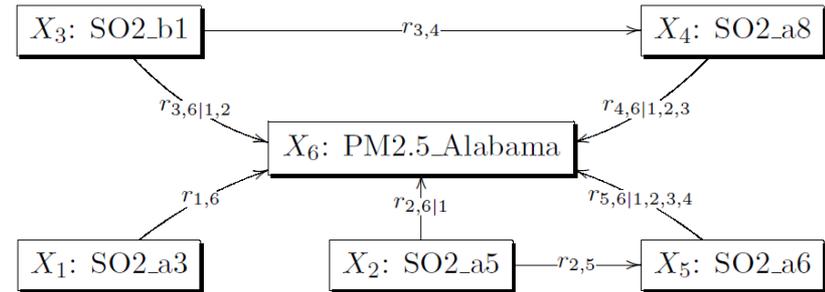
	V1	V2	V3	V4
V1	1	0	0	0,45
V2	0	1	0	0,16
V3	0	0	1	-0,62
V4	0	0	-1	1



- Bounds for CPE get shorter
- Not so much for RRC
- Which Method gives “better” estimates?
- Can experts estimate dependence?

# TO ANSWER THE QUESTIONS

- ▶ Let experts quantify models that we know
- ▶ Underlying assumptions hold
- ▶ SO<sub>2</sub> emissions and PM<sub>2.5</sub> concentrations
- ▶ See how well they can approximate the models
- ▶ M1 data M2 fictitious dependence



# RESULTS (INDIVIDUAL ESTIMATES)

$$|r_{i,j|D} - r_{i,j|D}^e|$$

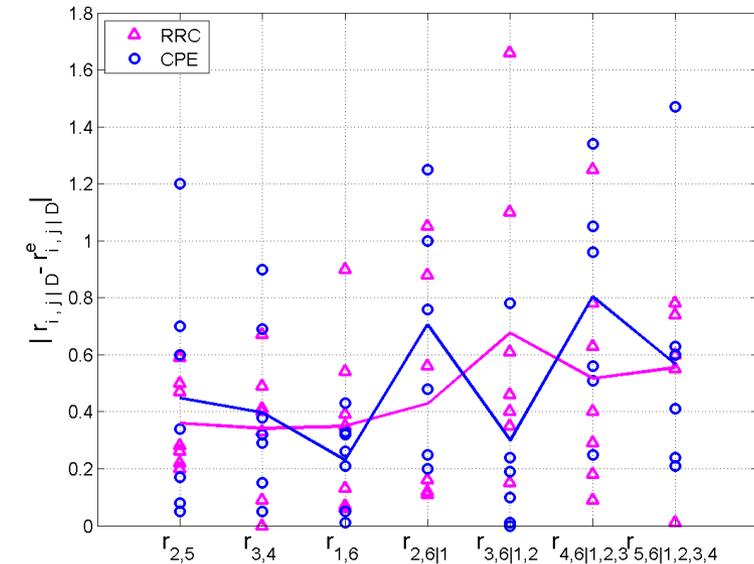
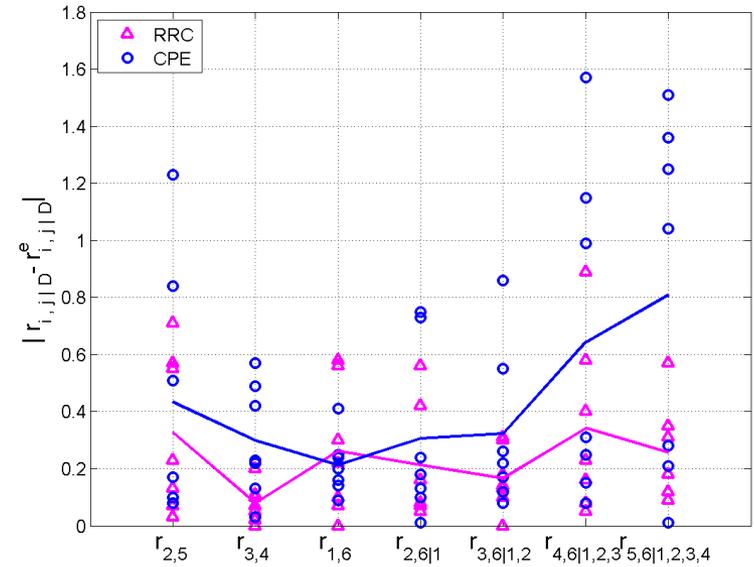
$$\delta_{M1RRC} = 0.23$$

$$\delta_{M1CPE} = 0.43$$

$$\delta_{M2RRC} = 0.46$$

$$\delta_{M2CPE} = 0.49$$

- › Mean difference half of other groups for
- › Ratios Rank Correlation with Data
- › Difference statistically significant
- › Doesn't say if individual experts can approximate the model of interest
- ›  $H_0: BN_e = BN_{true}$
- › Initial idea use det of corr. matrices





# DEPENDENCE CALIBRATION

- We have shown:
- $dCal = 1$  iff  $\Sigma_C = \Sigma_E$
- High dCal if calibration CM well approximated element wise
- Low dCal if high (Abs) correlations not well approximated
- Performance in assessing uncertainty and dependence don't correlate perfectly
- We must use other measures of performance for dependence
- Combinations based on dCal outperform individual opinions

$$H(f_C, f_E) = \iint_{[0,1]^2} \sqrt{\frac{1}{\sqrt{2}} (\sqrt{f_C(u, v)} - \sqrt{f_E(u, v)})^2} du dv$$

$$H_G(\Sigma_C, \Sigma_E) = \sqrt{1 - \frac{\det(\Sigma_C)^{1/4} \det(\Sigma_E)^{1/4}}{(1/2 \det(\Sigma_C) + 1/2 \det(\Sigma_E))^{1/2}}}$$

$$D = 1 - H$$

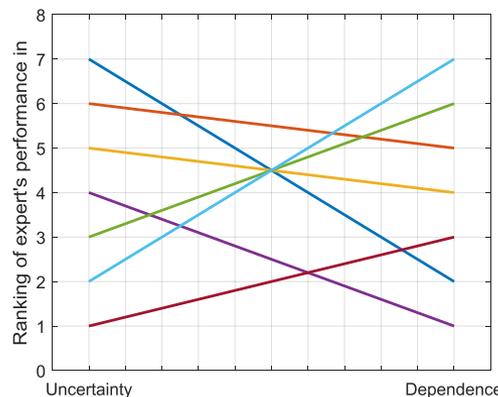
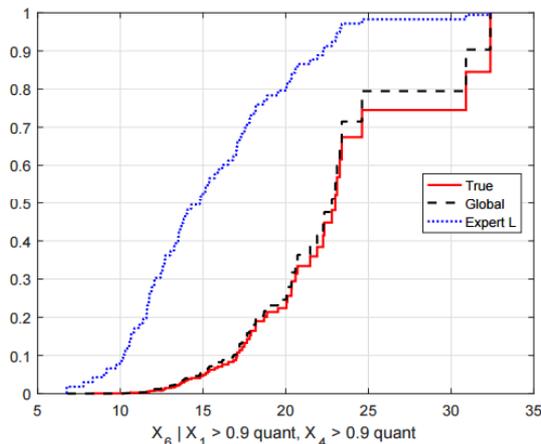


Fig. 1 Ranking of experts performance as uncertainty assessors and dependence assessors.

Group 2				
Id.	Calibr.	Inform.	$dCal$ M1RRC	$dCal$ M2CPE
D	0.0357	2.745	0.71	0.60
E	0.0063	1.497	0.51	0.32
F	0.7069	0.7571	0.12	0.09
G	l.o.	1.86	0.87	0.49
J	l.o.	2.49	0.32	0.17
L	0.0028	1.169	0.09	0.16
M	0.00131	3.84	0.75	0.32
Eq.	0.5503	0.3009	0.66	0.37
Gl.	0.7069	0.7571	0.95	0.60

Table 3: Calibration, Information and d-Calibration scores for air pollution NPNB experts.

# WHAT IF CORRELATION IS NOT SUFFICIENT?

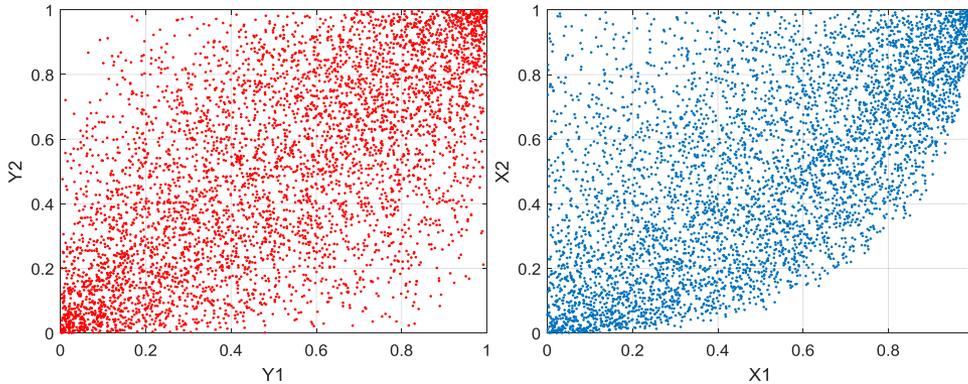


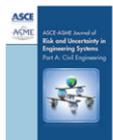
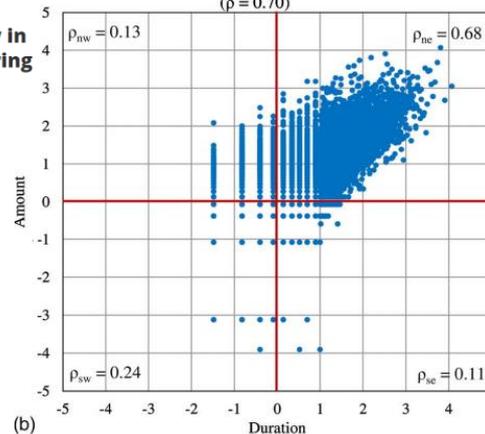
Fig. 3 Samples from a bivariate Gaussian copula ( $Y_1, Y_2$ ) and a copula with asymmetries ( $X_1, X_2$ ).

$$H(f_C, f_E) = \iint_{[0,1]^2} \sqrt{\frac{1}{2}} \left( \sqrt{f_C(u, v)} - \sqrt{f_E(u, v)} \right)^2 dudv$$

## Characterization of Precipitation through Copulas and Expert Judgement for Risk Assessment of Infrastructure

- Started using technique for problems where asymmetries exist.
- Can exp. identify asymmetries?
- Evaluate expert's ability to recognize asymmetries?
- Combine exp. (disagree) w.r.t. possible asymmetries?
- (Dis)agreement grows with increased dimensionality?
- Complexity in combination grows with increased complexity?
- Similar conclusions

Rain amount and duration, De Bilt, 1951-2013  
( $\rho = 0.70$ )



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# CONCLUSIONS

- › EXPERTS CAN ESTIMATE DEPENDENCE
- › DEPENDENCE ELICITATION IS ESSENTIALLY DIFFERENT THAN UNCERTAINTY ELICITATION
- › USUAL MEASURES FOR UNCERTAINTY ELICITATION ARE NOT SUFFICIENT.
- › EVALUATE EXPERT PERFORMANCE WITH APPROPRIATE MEASURES
- › COMBINATIONS BASED ON DCAL OUTPERFORM INDIVIDUAL OPINIONS
- › STILL MUCH TO BE DONE