

Generalised Co-variation for Sensitivity Analysis in Bayesian Networks



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 - ▷ in general
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Sensitivity analysis in Bayesian networks

- **Sensitivity analysis:** a standard technique for studying effect of changes in model parameters on model output
- **in Bayesian Networks:** output probabilities are simple, multi-linear functions of network parameters (CPT entries)

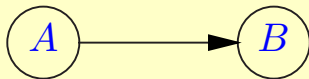
Example:

A probability $\Pr(v)$ as a function of 2 network parameters x_1, x_2 :

$$f_{\Pr(v)}(x_1, x_2) = c^{11} \cdot x_1 \cdot x_2 + c^{10} \cdot x_1 + c^{01} \cdot x_2 + c^{00}$$

- ▶ posterior \rightarrow quotient
- ▶ **assumption:** (proportional) co-variation of other entries from same distribution

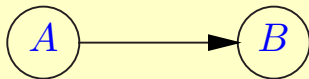
Co-variation in 1-way analysis



Varying a single parameter for a binary-valued variable:

	a_1	a_2		a_1	a_2
b_1	0.8	0.4	\implies	x	0.4
b_2	0.2	0.6		$1 - x$	0.6

Co-variation in 1-way analysis



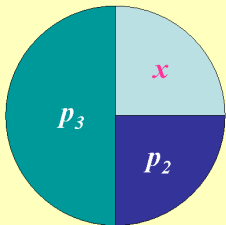
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Varying a single parameter for a multi-valued variable:

	a_1	a_2		a_1	a_2
b_1	0.5	0.1	\implies	x	0.1
b_2	0.2	0.5		} $1 - x$	0.5
b_3	0.3	0.4			b_3

Proportional co-variation



$$p_2 = 0.25$$

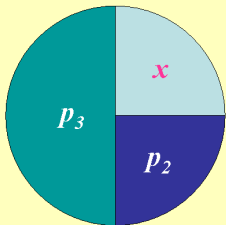
$$p_3 = 0.50$$

$$1 - x$$

$$0.75$$

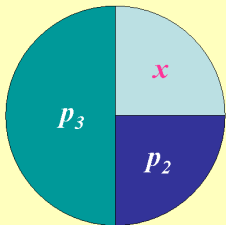
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Proportional co-variation

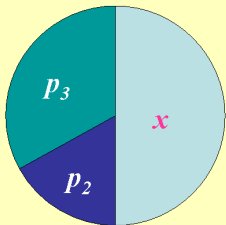


$$p_2 = 0.25 = 0.33 \cdot 0.75$$
$$p_3 = 0.50 = 0.67 \cdot 0.75$$

Proportional co-variation

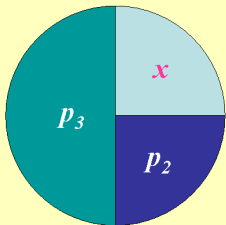


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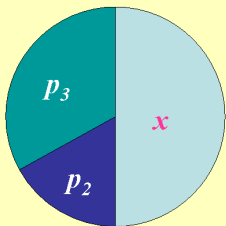


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Proportional co-variation



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$$p_2 = 0.17 = 0.33 \cdot 0.50$$
$$p_3 = 0.33 = 0.67 \cdot 0.50$$

Motivation I

Why use **proportional** co-variation?

- ▶ standard approach
- ▶ assumed by sensitivity functions, algorithms & properties
- ▶ seems sensible
- ▶ works with any parameter
- ▶ **optimal**

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- ▶ works with any parameter
- ▶ **optimal**:

The **CD-distance** between the original distribution \Pr and the new distribution \Pr^*

$$D(\Pr, \Pr^*) = \ln \max_w \frac{\Pr^*(w)}{\Pr(w)} - \ln \min_w \frac{\Pr^*(w)}{\Pr(w)}$$

is smallest under a proportional co-variation scheme
[Chan & Darwiche, 2002]

▶ ...

Motivation II

Why use an **alternative** co-variation scheme?

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- ▶ do functions, algorithms & properties depend on scheme?
- ▶ is CD-distance really optimal?

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$n > 1$ simultaneous parameter changes can result in a smaller CD-distance [Chan & Darwiche, 2004]

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- ▷ **this is unknown, and not obvious. . .**

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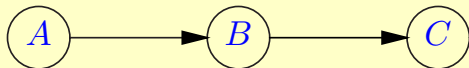
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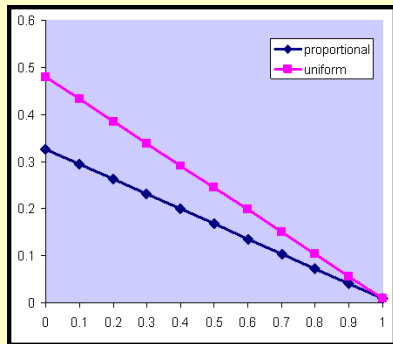
- ▷ **again smallest under proportional co-variation ?**
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- ▶ who cares about CD-distance? 😊
- ▶ why *minimise* 'disturbance' in a sensitivity analysis?
- ▶ ...

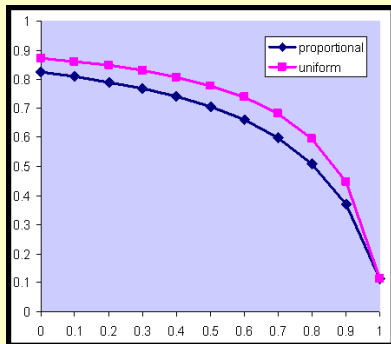
Examples



Parameter $x = p(b_1 | a)$ with $x_0 = 0.2$ is varied in steps of 0.1(!)

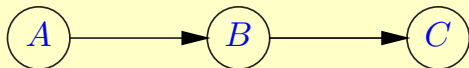


Output of interest:
 $\Pr(a, c)$ with $p_0 = 0.26$

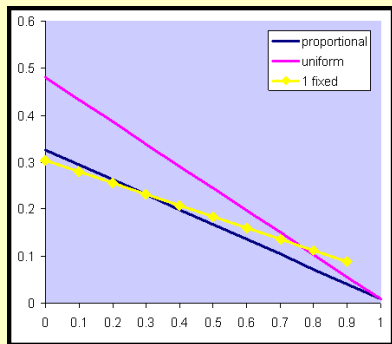


Output of interest:
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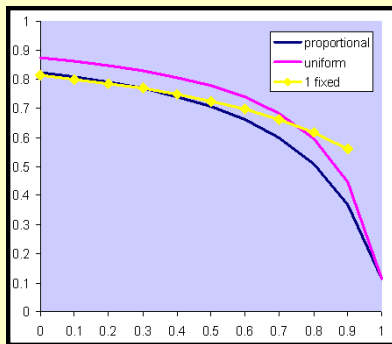
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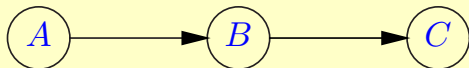


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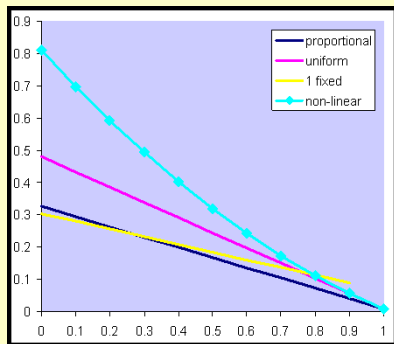


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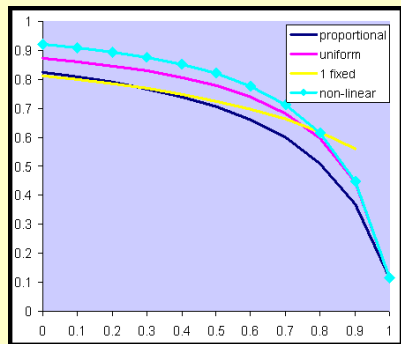
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Contributions

- We provide generalised formulas for the sensitivity function
 - ▷ which explicitly incorporate the co-variation scheme
 - ▷ for both 1-way and n -way functions
 - ▷ also when parameters are fixed
 - ▷ standard form is preserved for co-variation schemes
linear in x
- We provide generalised formulas for the CD-distance
 - ▷ which show that optimality of the proportional scheme is not obvious in the n -way case
 - ▷ and prove a lowerbound for single CPT co-variation under the proportional scheme