Explainable Bayesian Networks via Natural Language Explanations and Interactive Visualisation

Miruna Clinciu, Herbert Lau and Michael John Williams
Schlumberger Cambridge Research, Cambridge, UK

MClinciu@slb.com

Overview

Bayesian Networks represent an important modeling technique that can deal with **uncertainty** in knowledge-based systems.

However, complex BNs draw diffidence and criticism due to insufficient transparency and information on how the results were reached.

Key Takeaways:

- Explainability
- Building Trust for expert and non-expert users
- Evaluation of NL Explanations
- Increasing Transparency in an industrial setting

METHODS

Interactive Visualisation

We added three "new ingredients" to a simple acyclic graph (DAG):

- 1. A specific colour code for inferred nodes
- For each edge between nodes we added an infographic icon that will provide more information
- 3. Each node will receive a shape similar to those of a process flowchart diagram, therefore familiar to many engineering disciplines.

NL Explanations

We provided three types of NL Explanations, produced by expert annotators, taking into consideration the technical aspects.

- 1. Causal NL Explanations (counterfactuals)
- An explanation that provides an overall explanation of how a decision was achieved.
- 3. NL Explanations are attached to each edge, that will provide a technical reason.

Schlumberger

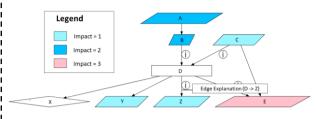


Figure 1. Interactive visualisation of a Bayesian Network, where colours indicate the relative importance of several key influences on the final decision; shapes similar to those of a process flowchart (parallelogram: Input, rectangle: inferred and diamond: Output to decision).

FUTURE WORK

In a study, these NL Explanations and different visualisation techniques will be rated in terms of informativeness, clarity, effectiveness and trust.