#### Knowledge and data engineering with Bayesian networks (a homework solution guide) Péter Antal

# Outline

- Tasks in the homework
- Knowledge enginering steps
  - Importance of ordering
  - Canonical models
    - Conditional probability decision trees

#### Homework

- Guide
- Tool
  - <u>http://redmine.genagrid.eu/projects/bayescub</u> edownload/wiki/Wiki
- Manual
- List of illustrative domains

# Goal of the homework

# To demonstrate and practice this multifaceted nature of Bayesian networks.

- As a probabilistic logic knowledge base, it provides a coherent framework to represents beliefs (see Bayesian interpretation of probabilities).
- As a decision network, it provides a coherent framework to represent preferences for actions.
- As a dependency map, it explicitly represents the system of conditional independencies in a given domain.
- As a causal map, it explicitly represents the system of causal relations in a given domain.
- As a decomposable probabilistic graphical model, it parsimoniously represents the quantitative stochastic dependencies (the joint distribution) of a domain and it allows efficient observational inference.
- As an uncertain causal model, it parsimoniously represents the quantitative, stochastic, autonomous mechanisms in a domain and it allows efficient interventional and counterfactual inference.

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# Obligatory and optional subtasks

- The minimal level contains the following subtasks (10 point):
  - Select a domain, select candidate variables (5-10), and sketch the structure of the Bayesian network model.
  - Consult it.
  - Quantify the Bayesian networks.
  - Evaluate it with global inference and "information sensitivity of inference" analysis.
  - Generate a data set from your model.
  - Learn a model from your data.
  - Compare the structural and parametric differences between the two models.
- Optional tasks:
  - Analyse estimation biases (5 point).
  - Investigate the effect of model uncertainty and sample size on learning: vary the strength of dependency in the model (increase underconfidence to decrease information content) and sample size and see their effect on learning (10 point).

#### Consultation

The preliminary approval of your planned homework is mandatory!

#### Documentation

Domain description.	10-100 words	
Variable definitions, with definitions	<20 words/variable	
of their values.		
Structure of the Bayesian network.	Explain the (preferably) causal order of the variables	
	and interesting independencies in your model. 50-500	
	words + figure(s).	
Quantify the Bayesian networks.	Illustrate your estimation in your model. 50-200 words	
	+ table(s)/figure(s).	
Evaluate it with global inference and	20-100 words + table(s)/figure(s).	
"information sensitivity of inference"		
analysis.		
Compare the structural and	50-200 words.	
parametric differences between the		
constructed and learnt models.		
Analyse estimation biases.	250-500 words + table(s)/figure(s).	
Investigate the effect of model	500-1000 words + table(s)/figure(s).	
uncertainty and sample size on		
learning.		

The overall documentation can be 3-5 pages (minimal) or 5-10 pages (full).

# Submission

- After consultation(!)
- the model XML with its documentation should be sent by email to your consulent.
- Deadlines:
  - Soft: before the last week of the semester (5th of December)
  - Hard: before the end of the semester (15th of December).

# Subtasks: importance of causality

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# Subtasks: canonical models

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# Noisy-OR

Noisy-OR distributions model multiple noninteracting causes

- 1) Parents  $U_1 \ldots U_k$  include all causes (can add leak node)
- 2) Independent failure probability  $q_i$  for each cause alone

 $\Rightarrow P(X|U_1 \dots U_j, \neg U_{j+1} \dots \neg U_k) = 1 - \prod_{i=1}^j q_i$ 

Cold	Flu	Malaria	P(Fever)	$P(\neg Fever)$
F	F	F	0.0	1.0
F	F	Т	0.9	0.1
F	Т	F	0.8	0.2
F	Т	Т	0.98	$0.02 = 0.2 \times 0.1$
Т	F	F	0.4	0.6
Т	F	Т	0.94	$0.06 = 0.6 \times 0.1$
Т	Т	F	0.88	$0.12 = 0.6 \times 0.2$
Т	Т	Т	0.988	$0.012 = 0.6 \times 0.2 \times 0.1$

Number of parameters linear in number of parents

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#### Decision trees, decision graphs



Decision tree: Each internal node represent a (univariate) test, the leafs contains the conditional probabilities given the values along the path. Decision graph: If conditions are equivalent, then subtrees can be merged. E.g. If (Bleeding=absent,Onset=late) ~ (Bleeding=weak,Regularity=irreg) A.I.: BN homework guide

# Subtasks: sensitivity of inference

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# Subtasks: learn model

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# Subtasks: estimation bias

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  - Consult it.
  - Quantify the Bayesian networks.
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  - Generate a data set from your model.
  - Learn a model from your data.
  - Compare the structural and parametric differences between the two models.
- Optional tasks:
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# Subtasks: effect of model uncertainty and sample size on learning

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  - Consult it.
  - Quantify the Bayesian networks.
  - Evaluate it with global inference and "information sensitivity of inference" analysis.
  - Generate a data set from your model.
  - Learn a model from your data.
  - Compare the structural and parametric differences between the two models.
- Optional tasks:
  - Analyse estimation biases (5 point).
  - Investigate the effect of model uncertainty and sample size on learning: vary the strength of dependency in the model (increase underconfidence to decrease information content) and sample size and see their effect on learning (10 point).

# Summary

- The homework takes you through real stages of knowledge engineering and machine learning:
  - Select a domain, create variables (5-10), and specify structure.
  - Quantify the Bayesian network.
  - Analyse estimation biases
  - Evaluate it with "information sensitivity of inference" analysis.
  - Generate a data set from your model.
  - Learn a model from your data.
  - Compare the structural and parametric differences between the two models.
  - Investigate the effect of model uncertainty and sample size on learning.