## Hugin & Weka for Learning from Data

Using Bayesian Networks

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## Learning from Data?

Predicting Probability distribution of unknown variables from data/cases

# Resulting learning model can be used as classifier

**Knowledge discovery** 

### Classification Task

Classifying variable y = x0 called the class variable given set of variables X=x1,x2...xk called attribute variables Classifier is learned from a dataset D consisting of samples over (x,y) on network Bx over probability distribution U

## Learning using Hugin

Uses EM (Estimation Maximization) Algorithm (Batch learning) This is used only when structure is available Experience table must be provided for nodes

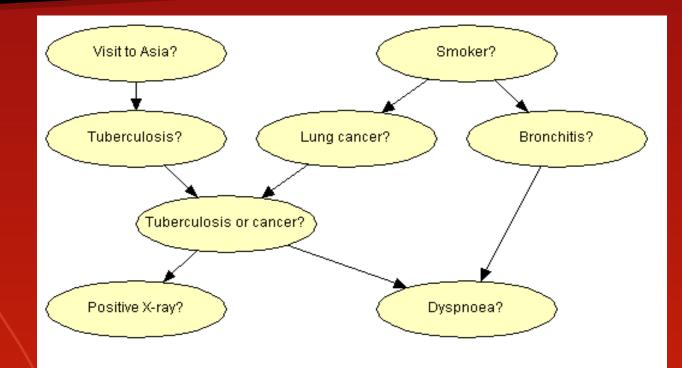
whose conditional probabilities are to inferred



Performs number of iteration on cases Computes log-likelihood and attempts to maximize it Stops when two successive log-likelihood is

less than tolerance

# Learning on Chest Clinic data



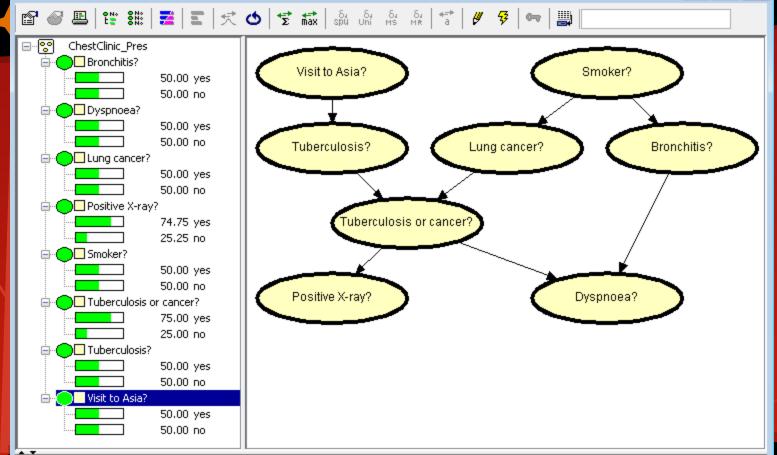


No prior knowledge on any distribution

#### Set probability distribution to 1 except for "Tuberculosis or Cancer"

Also experience value to 10 (or some low value)

🙀 Class: ChestClinic\_Pres



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#### Data set (asia.data)

#### First line is header

#### Each record is a case

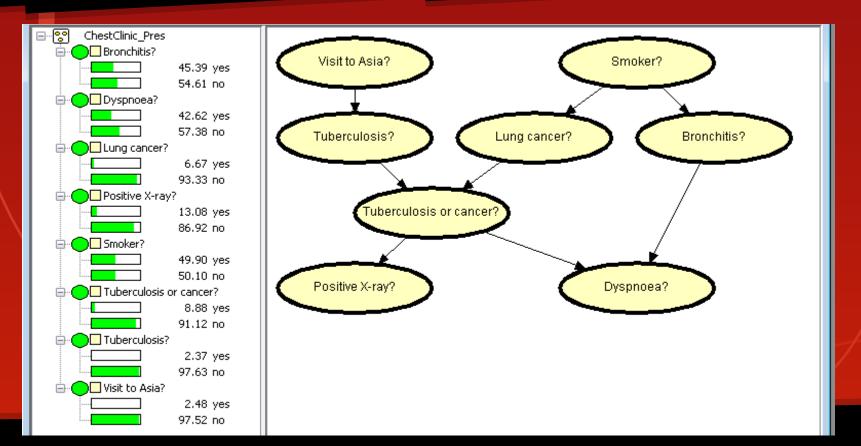
#### N/A = Not available

a 🔢	🔚 asia.dat 🗵				
:	E,T,L,S,A,D,B,X				
1	no,no,no,yes,no,yes,yes,no				
:	no,no,no,yes,no,no,yes,no				
1.1	no,N/A,no,no,no,yes,yes,no				
1	no,no,no,no,no,no,no				
	no,no,no,no,yes,yes,no				
1	no,no,no,yes,no,no,yes,no				
1	no,no,no,no,no,no,no				
1	no,no,no,yes,no,yes,no,no				
1	no,no,no,no,no,no,no				
1	no,no,no,yes,N/A,yes,yes,no				
1	no,no,no,yes,N/A,yes,yes,no				
1	no,no,no,yes,no,no,no,no				
1	no,no,no,yes,no,no,no,no				
1	yes, no, yes, yes, no, yes, yes, yes				
1	N/A,no,no,N/A,no,yes,yes,N/A				
1	no,no,no,no,yes,yes,no				
1	no,no,N/A,yes,no,no,no,no				
1	no,no,no,yes,no,no,no,no				
2	no,no,no,no,N/A,no,no				
2	no,no,N/A,N/A,no,yes,yes,no				
23	no,no,no,no,no,no,no				
23	yes, yes, no, yes, no, no, yes, yes				
2	yes, no, yes, yes, no, no, no, yes				
2	no no no no ves ves N/A				

## Run the Learning Algo

	yes		10
2		0.052183	
B Load	d Data File	× 817	
Fill Se	elect File Browse le Options Encoding UTF-8 Separator Symbol  Comma Semicolon Tab  Space Other Load Cancel	•	EM Learning Parameters for EM Learning Data File: Number of Iterations: 0 Tolerance: 1.0E-4 Select File View Data OK Cancel
		_	
	C Dysphoea?	<u>&gt;</u>	· ·

# Resulting Marginal Probabilities





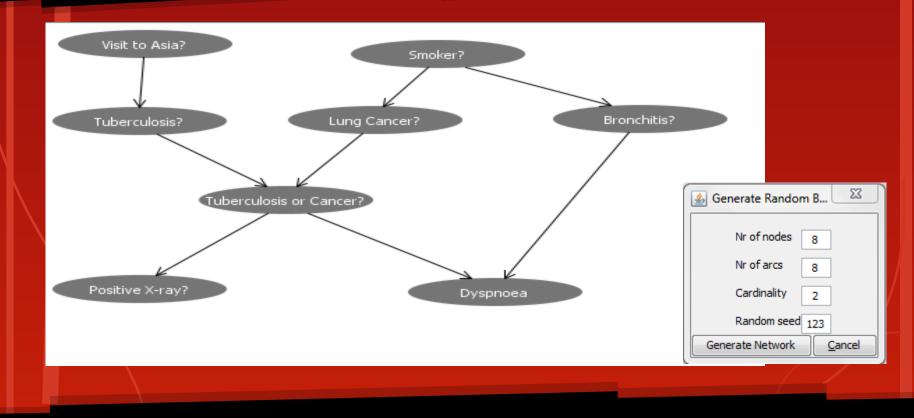
The resulting model with computed conditional probabilities can be used as classifier to predict the new unknown conditional probabilities

## Learning using WEKA

Collection of Machine Learning Algorithms for data mining tasks

Contains tools for data pre-processing, classification, clustering, visualization etc. Bayesian Network classifier and editor is one of them

## Creating Bayesian Network



All variables are discrete finite variables If continuous can convert to discrete using class filters.attribute.Discretize

No instances have missing values. If found can be filled by attribute.ReplaceMissingValues

Weka Explorer				
Preprocess Classify Cluster Associa	te Select attributes Visualize			
Classifier				
Choose BayesNet -D -B ChestClinic_weka_nw.xml -Q weka.classifiers.bayes.net.search.fixed.FromFileB -E weka.classifiers.bayes.net.estimate.Si				
Test options	Classifier output			
🔘 Use training set 🧔 weka.gu	i.GenericObjectEditor			
	ers.bayes.BayesNet Linic_weka_nw.xml -Q weka.			
Cross-validation     About				
Percentage split Bayes N	etwork learning using various search algorithms More			
	lity measures. Capabilities			
(Nom) E B	FFile ChestClinic_weka_nw.xml			
Start	ebug False			
Result list (right-click for	nator Choose SimpleEstimator			
11:57:02 - bayes.Bayes	Choose Simple: Scinator & weka.gui.GenericObjectEditor			
searchAlgo	rithm Choose FromFile -B weka.classifiers.bayes.net.search.fixed.FromFile			
useAE	Tree False About			
	The FromFile reads the structure of a Bayes net from a file in Mo			
Open.	Save BIFF format.			
	•			
	Open Save OK Cance			
Chabur				
Status Interrupted	Log x 0			

#### ✦ Results

Preprocess	Classify	Cluster	Associate	Select attributes	Visualize	
Classe: Care						

#### . .

Test options	Classifier output			
<ul> <li>Use training set</li> <li>Supplied test set Set</li> <li>Cross-validation Folds 10</li> <li>Percentage split % 66</li> <li>More options</li> </ul>	LogScore Bayes: -26771.70353988872 LogScore BDeu: -26783.728842977427 LogScore MDL: -26799.98957920934 LogScore ENTROPY: -26730.912026419 LogScore AIC: -26745.91202641952			
	Time taken to build model: 3.4 sec	onds		
(Nom) X Start Stop	=== Stratified cross-validation == === Summary ===	=		
Result list (right-dick for options) 11:57:02 - bayes.BayesNet 12:02:17 - bayes.BayesNet 12:04:14 - bayes.BayesNet	Correctly Classified Instances Incorrectly Classified Instances Kappa statistic Mean absolute error Root mean squared error Relative absolute error Root relative squared error Total Number of Instances	9483 517 0.6676 0.0753 0.223 39.7663 % 72.5024 % 10000	94.83 % 5.17 %	E
	=== Detailed Accuracy By Class ===			

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## Pros & Cons of Hugin

#### **Pros:**

Can generate data with missing values based on available network

Can predict prob. distribution for all variables Cons:

No feature for handling missing values No feature to predict class of all the cases at once

## Pros and Cons of WEKA

#### Pros:

Feature to handle missing values Can predict class of unknown data sets all at once Cons: Can't generate data with missing values Can't predict probability distribution

## Useful Links & References

- 1. Bouckaert, Remco R. Bayesian network classifiers in weka. Department of Computer Science, University of Waikato, 2004
- Mark Hall, Eibe Frank, Geoffrey Holmes, Bernhard Pfahringer, Peter Reutemann, Ian H. Witten (2009); The WEKA Data Mining Software: An Update; SIGKDD Explorations, Volume 11, Issue 1
- 3. More information on Hugin tool
- 4. More information on WEKA tool

