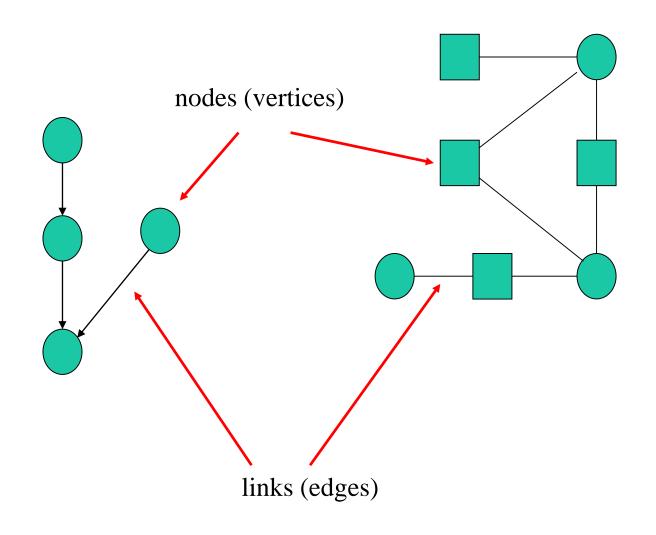
# Meeting 13: Graphical models



# Bayesian networks

Some general graphical model concepts:



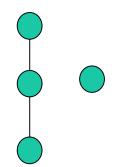
## A graph can be

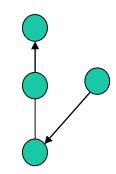
disconnected:

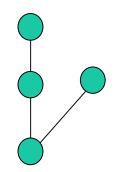
or *connected*:

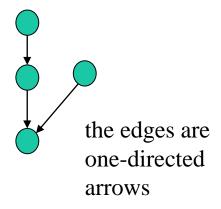
; undirected:

or directed:

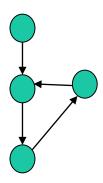






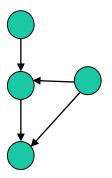


cyclic:



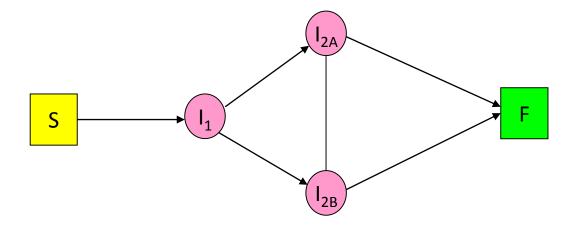
possible to start in one node and "come back"

or acyclic:



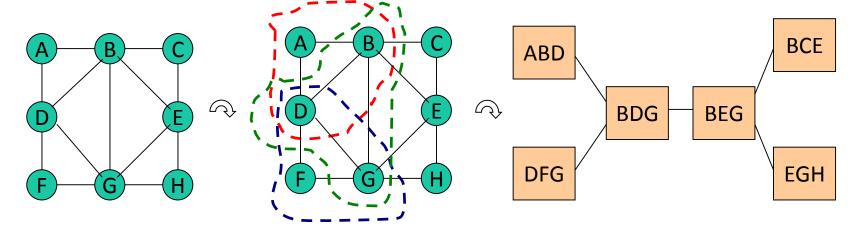
## Examples:

Transport routes:



Acyclic, but not completely directed

#### Junction trees:



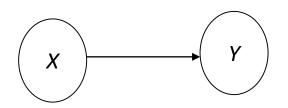
From 8 nodes to 6 nodes (Source: Wikipedia)

# Bayesian (belief) networks

A Bayesian network is a connected directed acyclic graph (DAG) in which

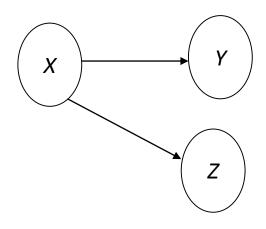
- the nodes (vertices) represent random variables
- the links (edges, arcs) represent direct *relevance* relationships among variables

## Examples:



This small network has two nodes representing the random variable *X* and *Y*.

The directed link gives a relevance relationship between the two variables that means  $Pr(Y = y \mid X = x, I) \neq Pr(Y = y \mid I)$ 



This network has three nodes representing the random variables X, Y and Z.

The directed links give relevance relationships that means

$$\Pr(Y = y \mid X = x, I) \neq \Pr(Y = y \mid I)$$

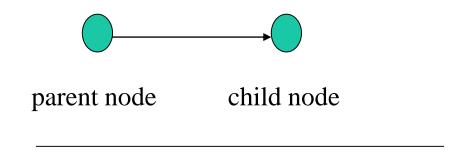
$$\Pr(Z = z | X = x, I) \neq \Pr(Z = z | I)$$

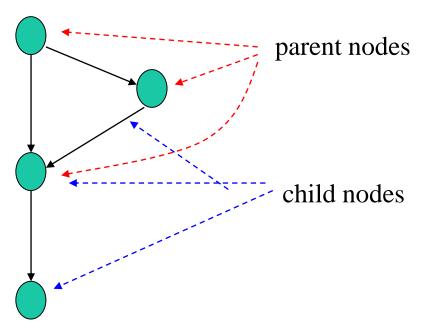
but also (as will be seen below)

$$Pr(Z = z | Y = y, X = x, I) = Pr(Z = z | X = x, I)$$

# Structures in a Bayesian network

There are two classifications for nodes: parent nodes and child nodes





Thus, a node can be solely a parent node, solely a child node *or* both!

## Probability "tables"

Each node represents a random variable.

This random variable has *either* assigned probabilities (nominal scale or discrete) or an assigned probability density function (continuous scale) for its states.

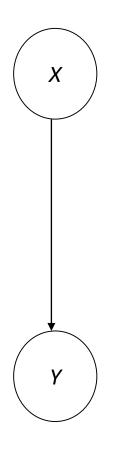
For a node that is *solely* a parent node:

The assigned probabilities or density function are conditional on background information only (may be expressed as unconditional)

For a node that is a child node (solely or joint parent/child):

The assigned probabilities or density function are conditional on the states of its parent nodes (and on background information).

## Example:



X has the states  $x_1$  and  $x_2$ 

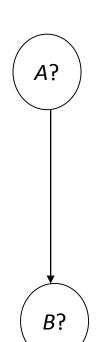
## Probability tables

X	Probabilities
$x_1$	$\Pr\left(X = x_1 \mid I\right)$
$x_2$	$\Pr\left(X = x_2 \mid I\right)$

Y has the states  $y_1$  and  $y_2$ 

		Probabilities			
X:		$x_1$	$x_2$		
<i>Y:</i>	<i>y</i> <sub>1</sub>	Pr $(Y = y_1   X = x_1, I)$	Pr $(Y = y_1   X = x_2, I)$		
	$y_2$	Pr $(Y = y_2   X = x_1, I)$	Pr $(Y = y_2   X = x_2, I)$		

## Example Dyes on banknotes (from previous lectures)



#### Two states:

A: "Dye is present"

 $\overline{A}$ : "Dye is absent"

<i>A</i> ?	Probabilities
A	0.001
$\overline{A}$	0.999

#### Two states:

B: "Result is positive"

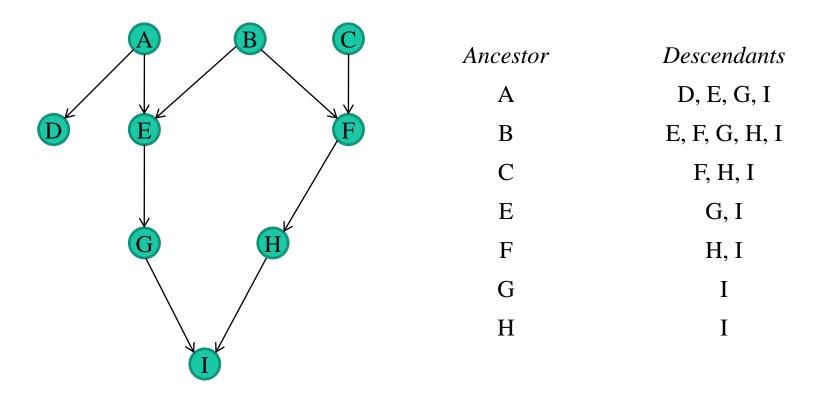
 $\overline{B}$ : "Result is negative"

		Probabilities		
A?:		A	$\overline{A}$	
<i>B</i> ?:	В	0.99	0.02	
	$\overline{B}$	0.01	0.98	

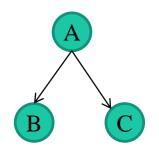
#### *More about the structure...*

### Ancestors and descendants:

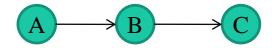
A node *X* is an *ancestor* of a node *Y* and *Y* is in turn a *descendant* of *X* if there is a unidirectional path from *X* to *Y* 



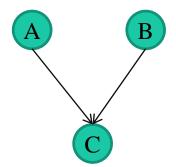
## Different connections:



diverging connection



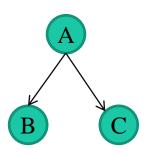
serial connection



converging connection

Conditional independence and d-separation

1) Diverging connection



There is a path between B and C even if it not unidirectional

→ B may be relevant for C (and vice versa)

However, if the state of A is known this relevance is lost.: The path is *blocked* 

→ B and C are *conditionally independent* given A

## Example

Let A be a random node with states

 $A_1 =$  "Willie is a cat"



 $A_2$  = "Willie is a parrot



Let B be a random node with states

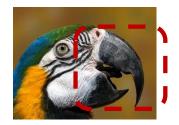
 $B_1$  = "Willie has four legs"

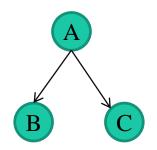
B<sub>2</sub> = "Willie has two legs"

Let C be a random node with states

 $C_1$  = "Willie has a beak"

 $C_2$  = "Willie has no beak"





Given B being equal to  $B_1$  the conditional probability of  $C_1$  is different (lower) than the conditional probability of  $C_1$  given B is equal to  $B_2$ .

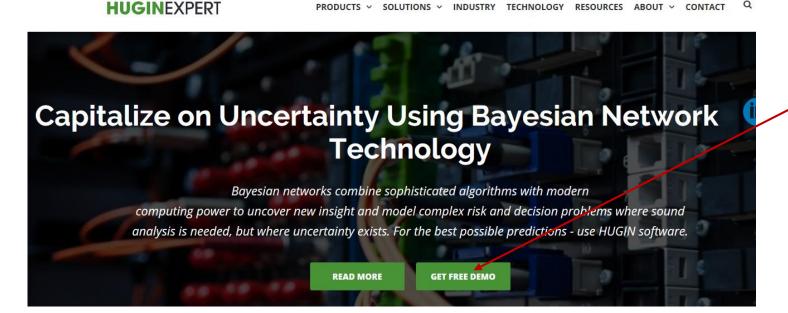
Hence, B is relevant for C and vice versa.

However, if A is instantiated to  $A_1$ , i.e. Willie is a cat, B and C are no longer relevant for each other if we reasonably assume that the number of legs a cat has cannot affect whether he has a beak or not.

#### Software

- Hugin (several types of commercial licenses available), Hugin Lite as demo version free of charge
- GeNIe (<a href="https://dslpitt.org/dsl/genie\_smile.html">https://dslpitt.org/dsl/genie\_smile.html</a>) used to be easy download freeware, but today it is more complicated
- Agena Risk (https://www.agenarisk.com/), trial version can be downloaded, otherwise commercial license needed
- ...several other

Hugin (www.hugin.com)



## Download the FREE HUGIN Lite demo

#### How to get started with **HUGIN** software

The best way to learn more about the HUGIN technology is to try it yourself. Download our free HUGIN Lite demo, a limited version of HUGIN Developer / Researcher.

The HUGIN Decision Engine in HUGIN Lite is available with interfaces for four different programming environments: C, C++, .NET, Java, and as an ActiveX-Server for Visual Basic.

It is prohibited to use the free HUGIN Lite for any other purpose than the demonstration of capabilities and proof of concept.

Go to our Technology Site to learn more about HUGIN Bayesian network technology.





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Address		
	Postal Code	
City		
	Country	
	Denmark	~
	Email *	
	Use of demo	
	Industrial application	
	Learn Hugin	
	Surfing around	~

**DOWNLOAD** 

#### Use of demo









Dear Anders Nordgaard,

Thank you for downloading the HUGIN Lite Evaluation.

The registration key required for installing this product and the links for download this product have already been sent to your email.

Please check your email to download it.

Thank you.

Thank you for downloading the HUGIN Lite demo.

The registration key required for installing this product is:

Email: anders.nordgaard@liu.se

<del>-</del>

Information to be used during installation

#### Download URLs:

Key: 2384

#### 32-bit:

http://download.hugin.com/pub/Licenses/8.8/HuginLiteR88.msi

#### 64-bit:

http://download.hugin.com/pub/Licenses/8.8/HuginLiteR88(x64).msi

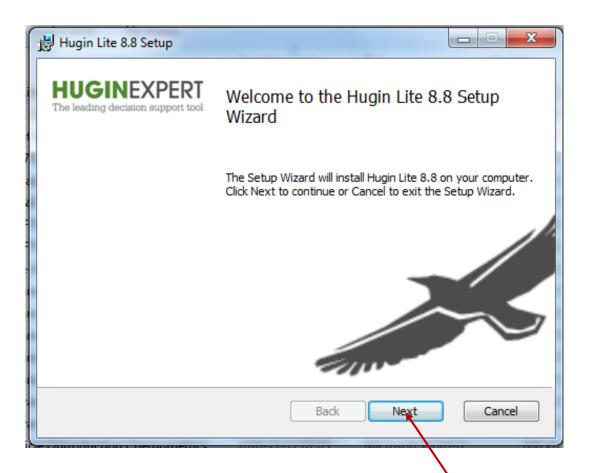
We hope you will find the product useful. If you need a quote for the full license, please contact sales@hugin.com

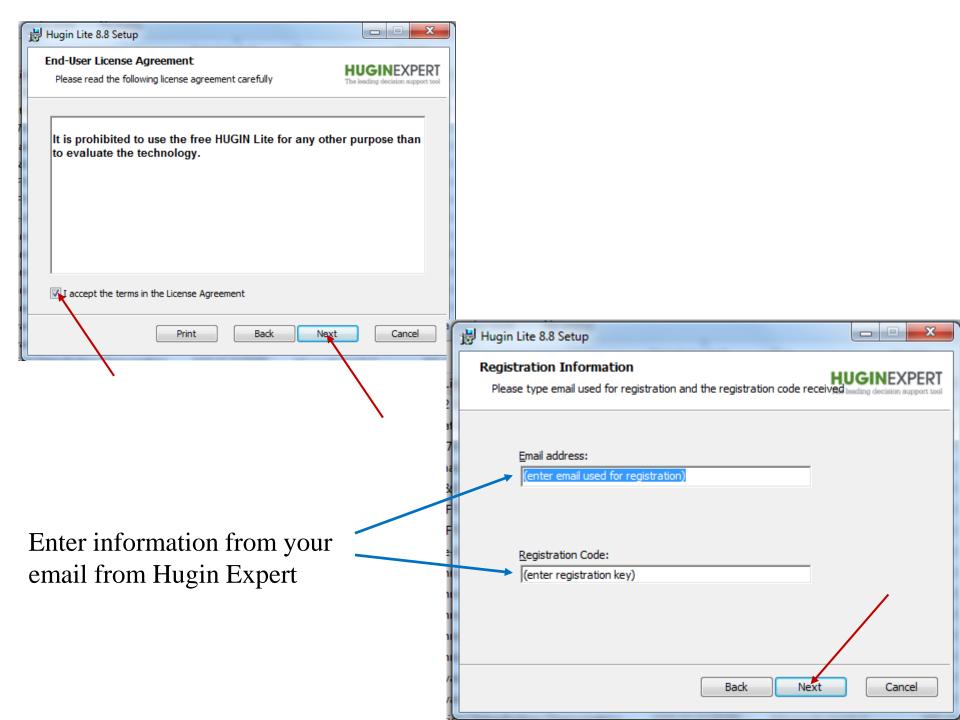
Best regards, HUGIN EXPERT A/S info@hugin.com

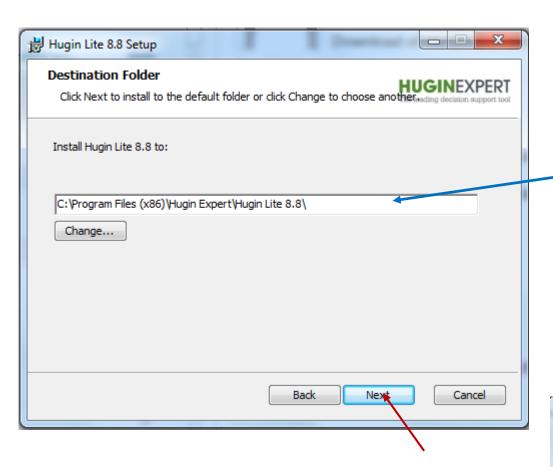


2010-10-22 16:25

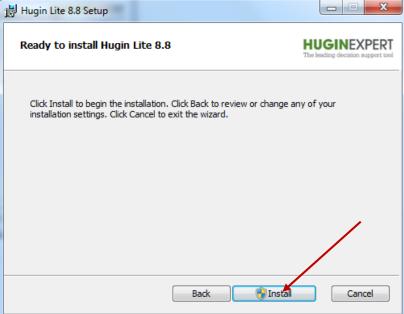
15 JR



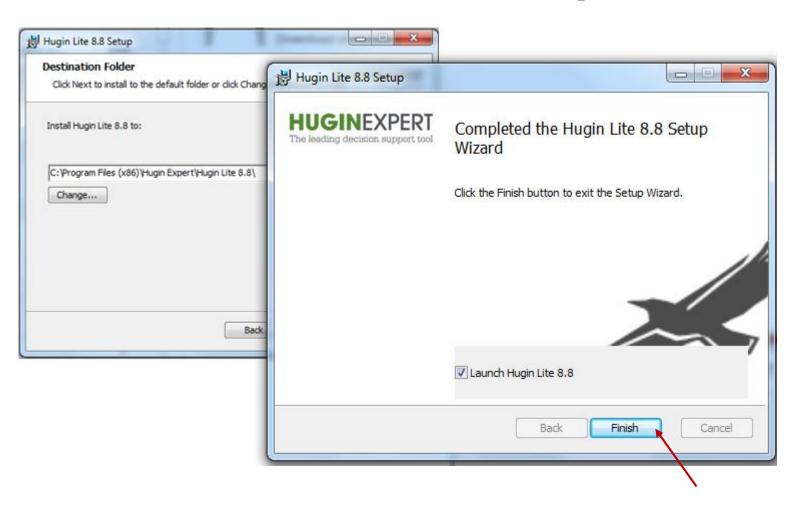




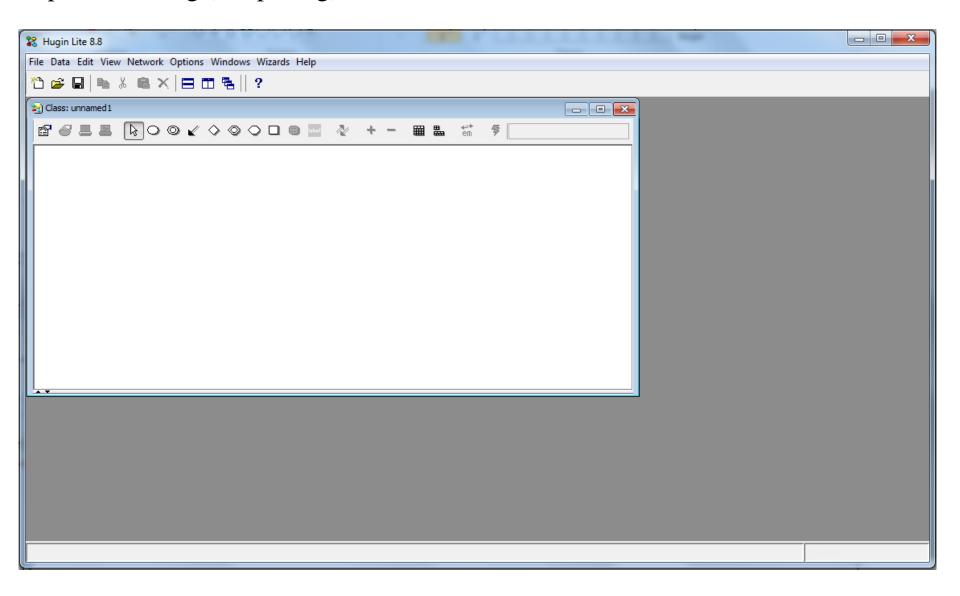
Use this or your own preferred folder



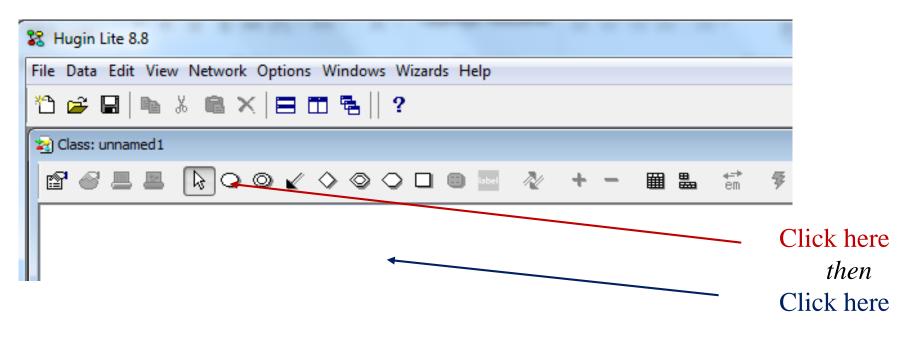
Installation takes less than one minute on most computers.

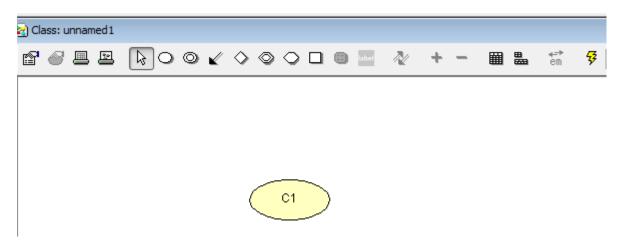


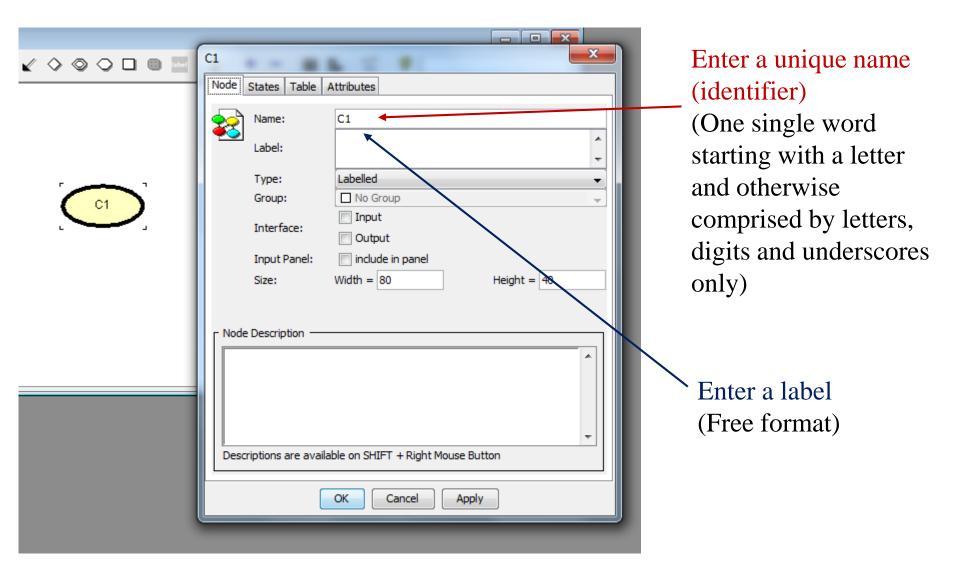
## Upon launching (or opening):



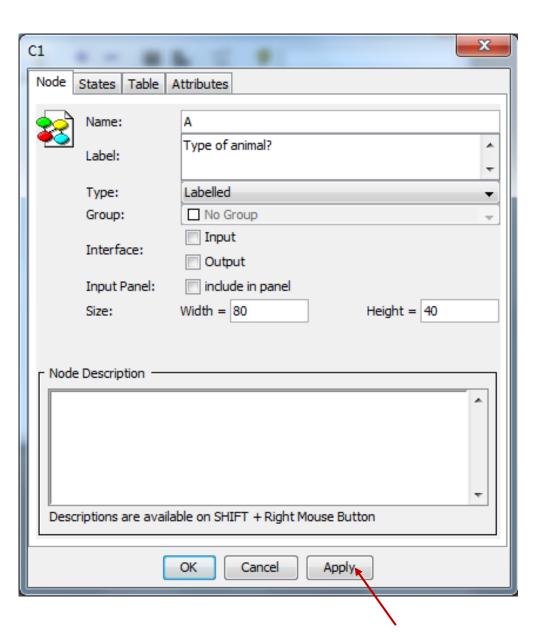
## Adding a chance node



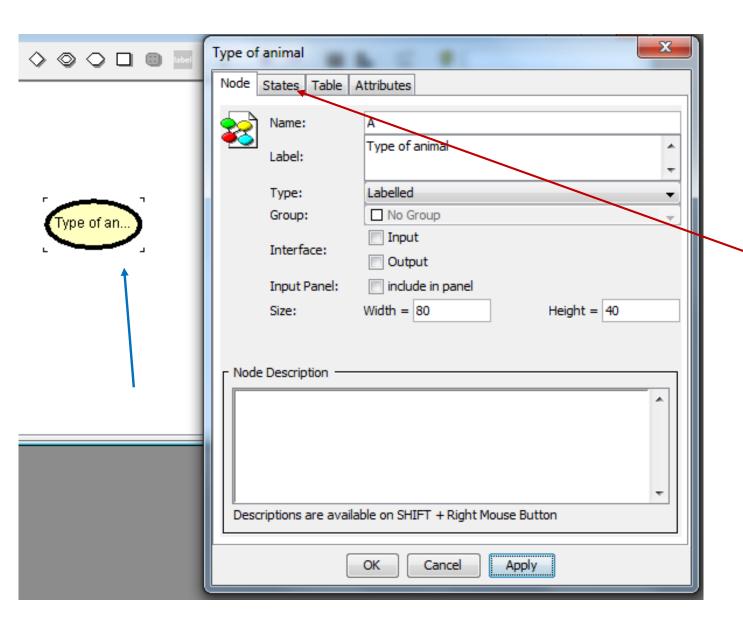




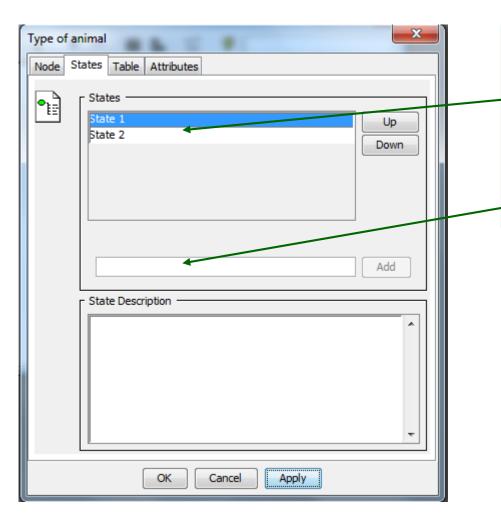
## ...for instance...



A<sub>1</sub> = "Willie is a cat" A<sub>2</sub> = "Willie is a parrot



Select tab "States"

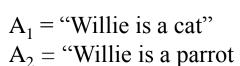


Two states per default

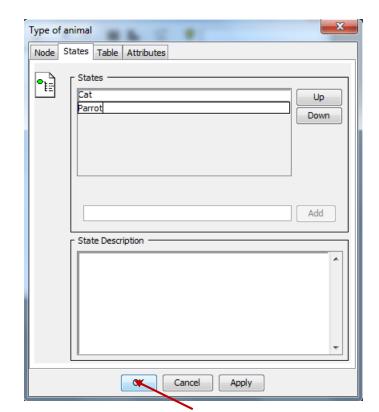
State names can be altered (double-click)

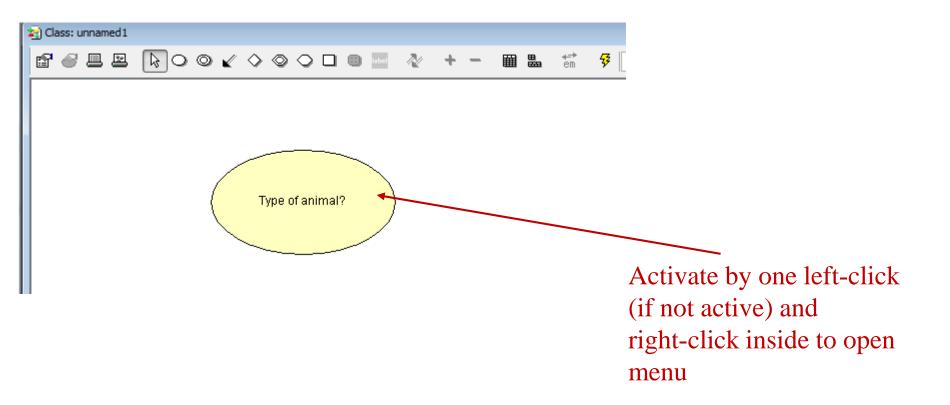
New states can be added

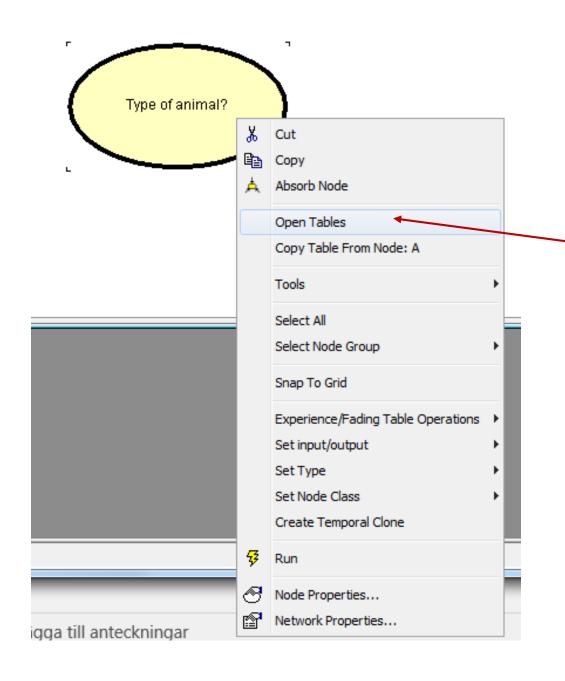
for instance...



No more selections needed in this dialogue.

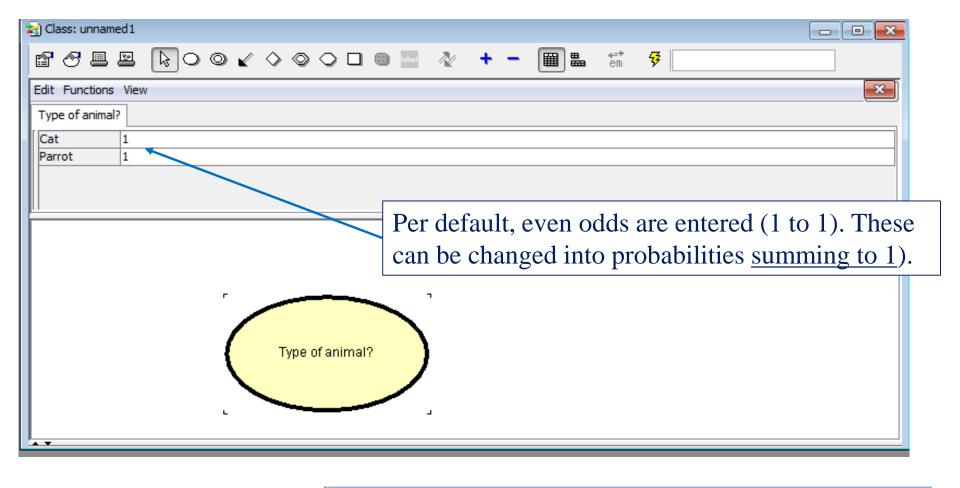


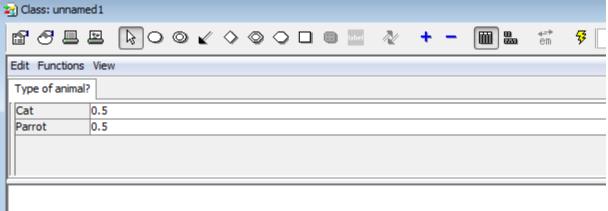




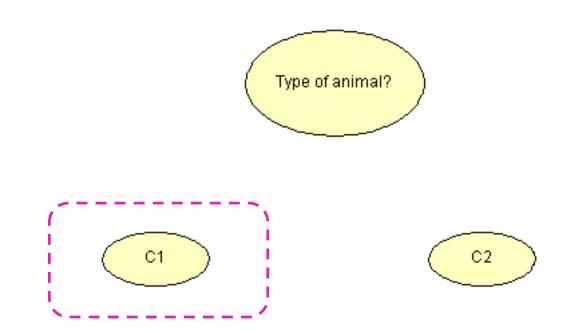
Select "Open Tables"

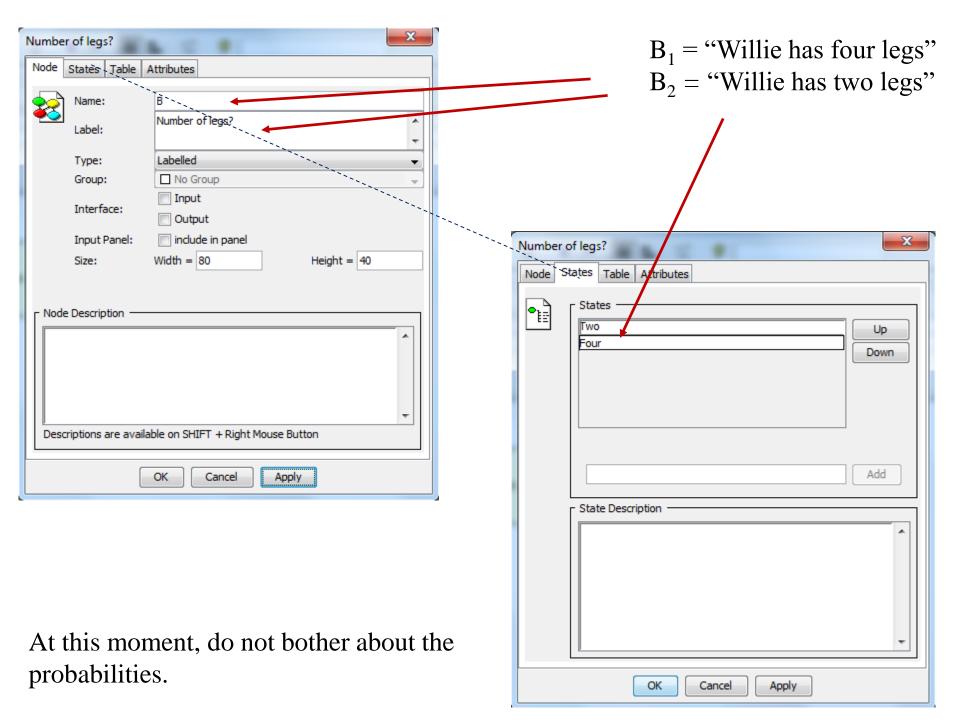
...might be followed by a warning with instructions about how a node table should be visible

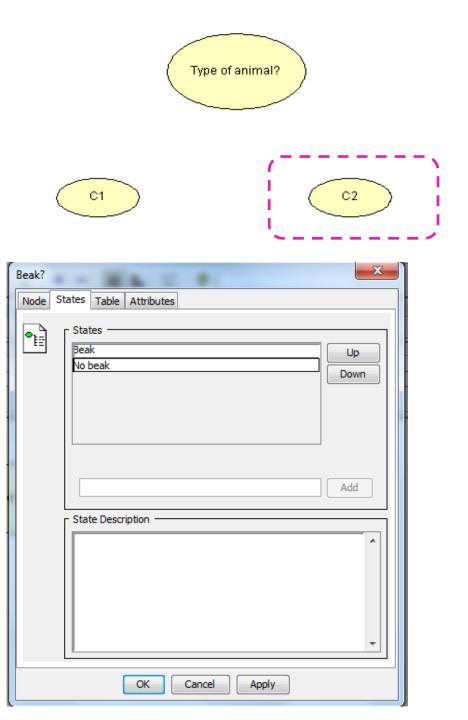




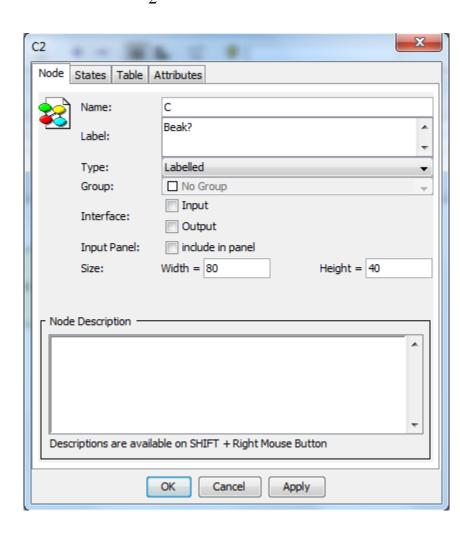
Add two more chance nodes...



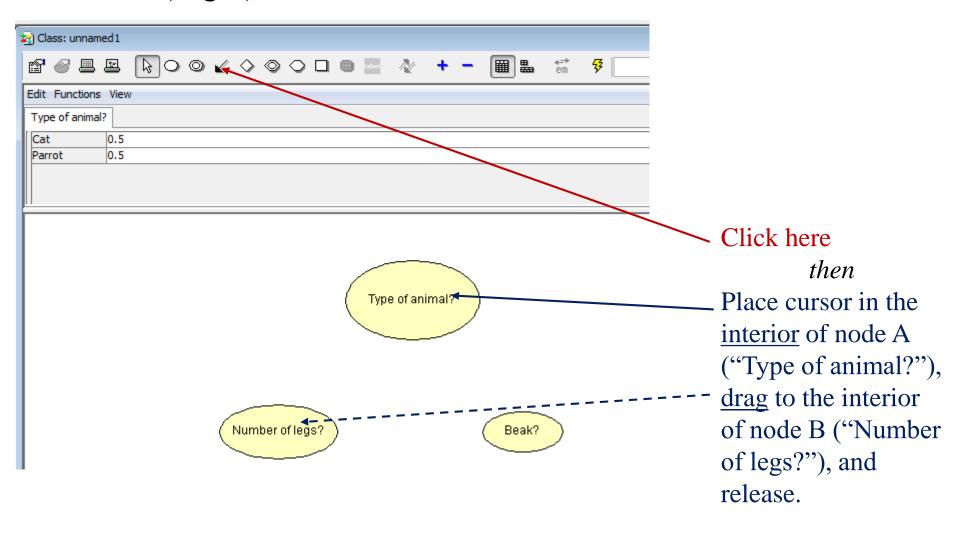




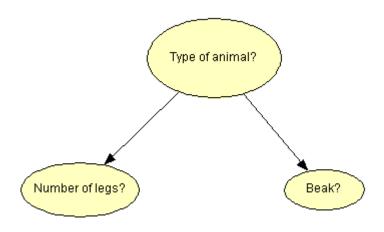
C<sub>1</sub> = "Willie has a beak"C<sub>2</sub> = "Willie has no beak"



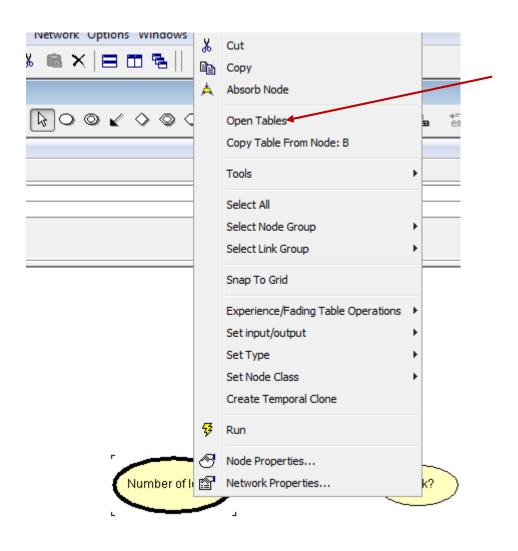
## Add links (edges)...

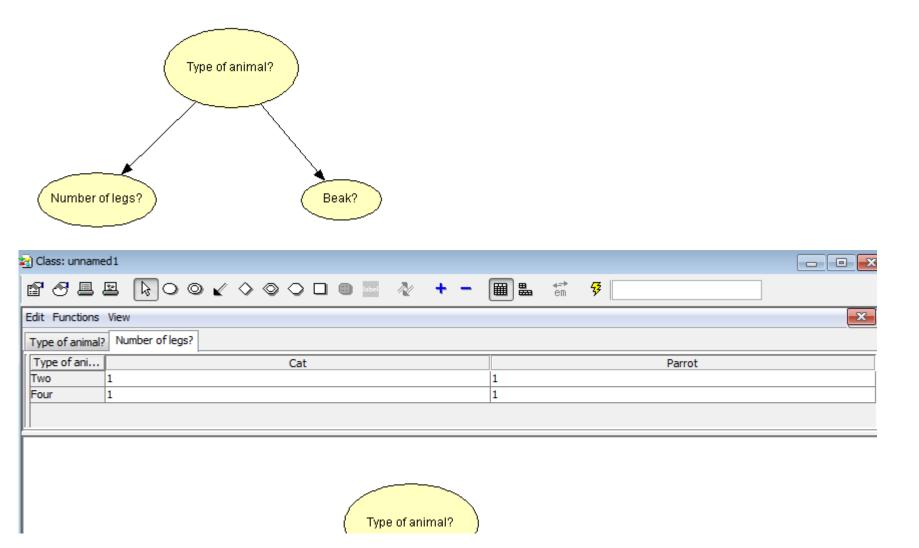


Repeat for link between A and C



Now, activate node B (Number of legs?), right-click and select Open Tables

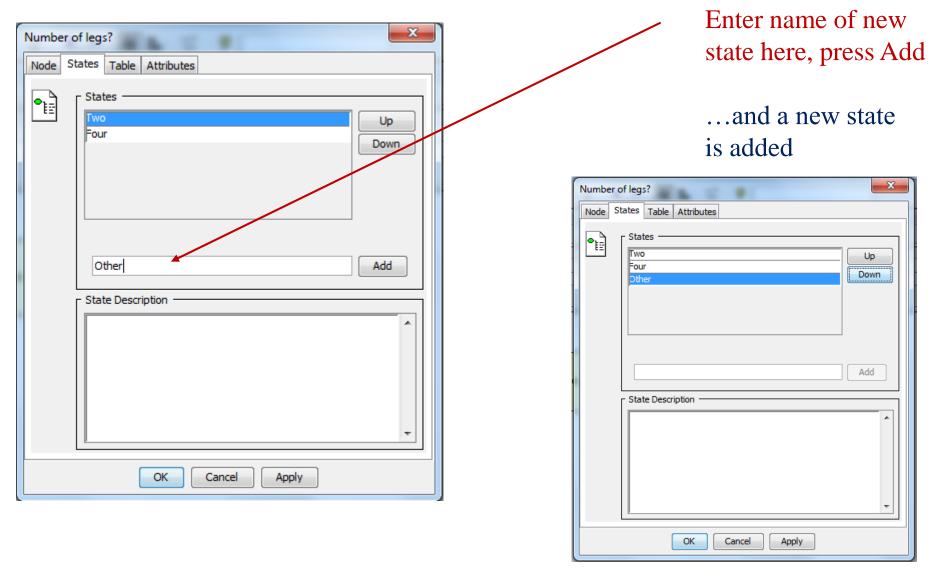




The probability table of node B now has probabilities conditional on the state of node A. The are all set to 1 per default, and will each be treated as 0.25.

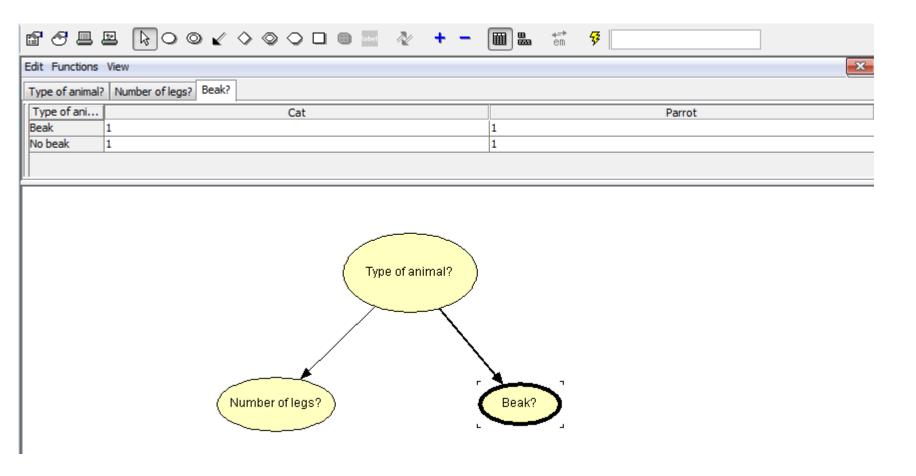
Are the states sufficiently many?

# Add a state. Double-click on node, select tab States



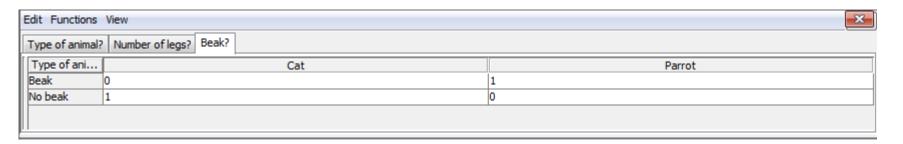
The order of the states can be changed by pressing Up and/or Down

Now, open the table for node C (Beak?) as well.



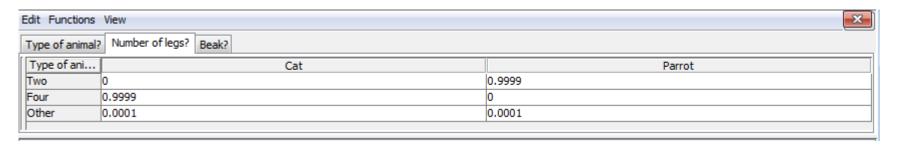
We can now enter our probabilities into the tables.

### Node C:



## Reasonable?

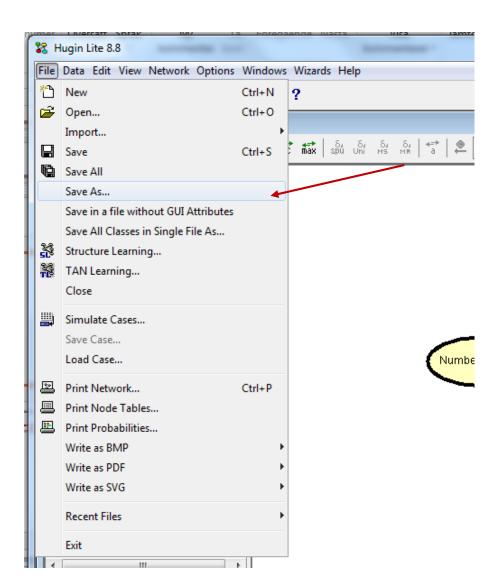
# Node B:

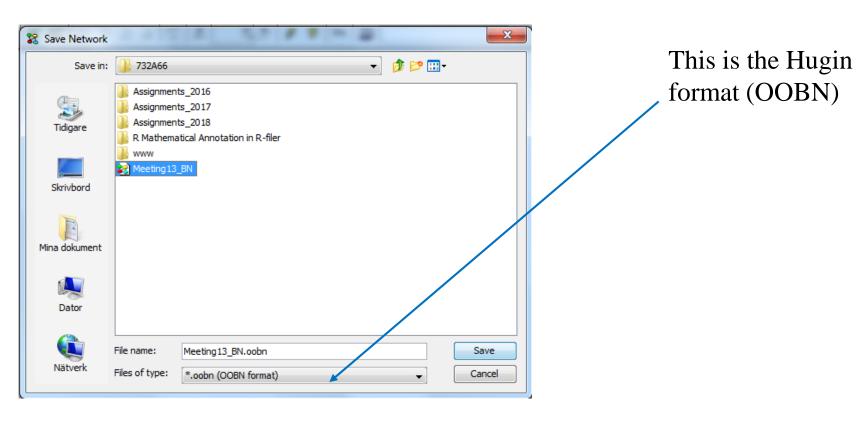


### Reasonable?

At this moment (or even earlier) it is wise to save the constructed network.

Open the File menu of the GUI and select Save As...





\*.oobn (OOBN format)

All Files

\*.oobn (OOBN format)

\*.net (net format)

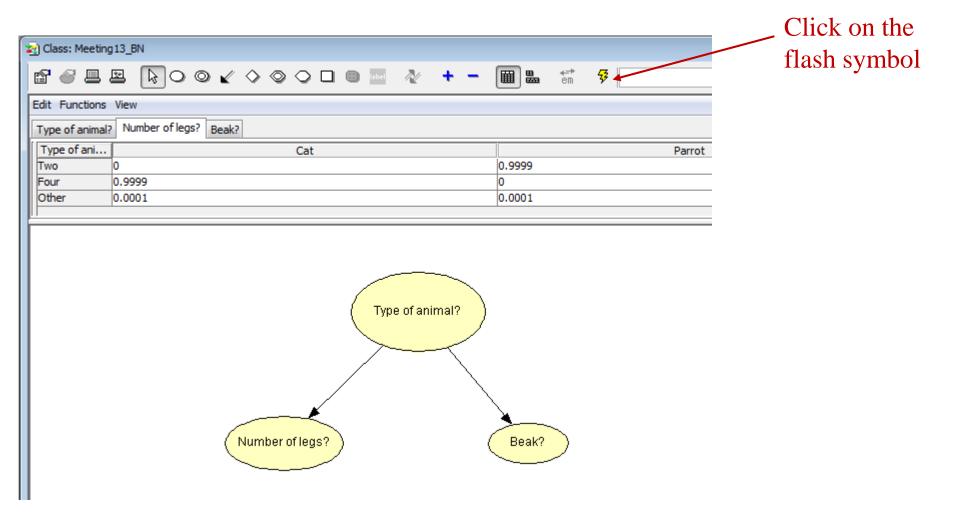
\*.hkb (net format)

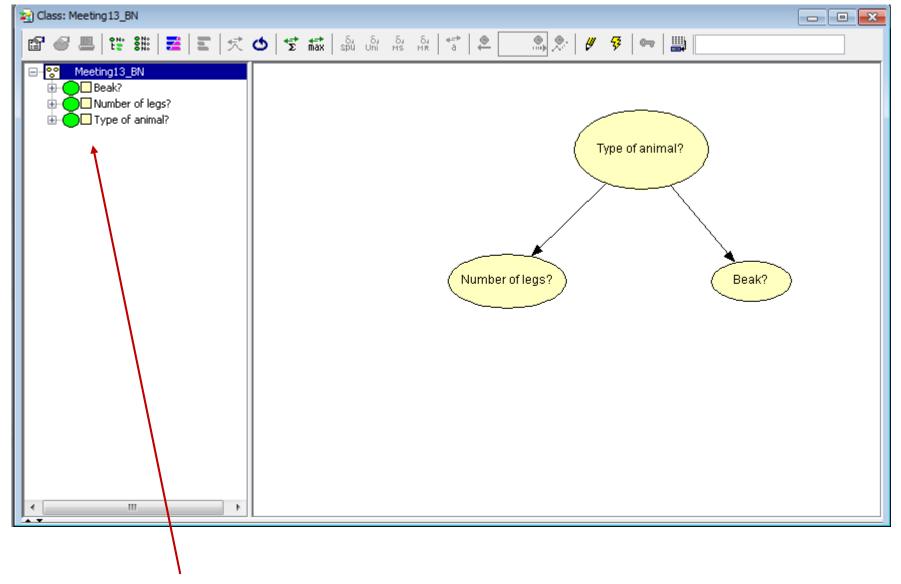
Password protected hkb

File name:

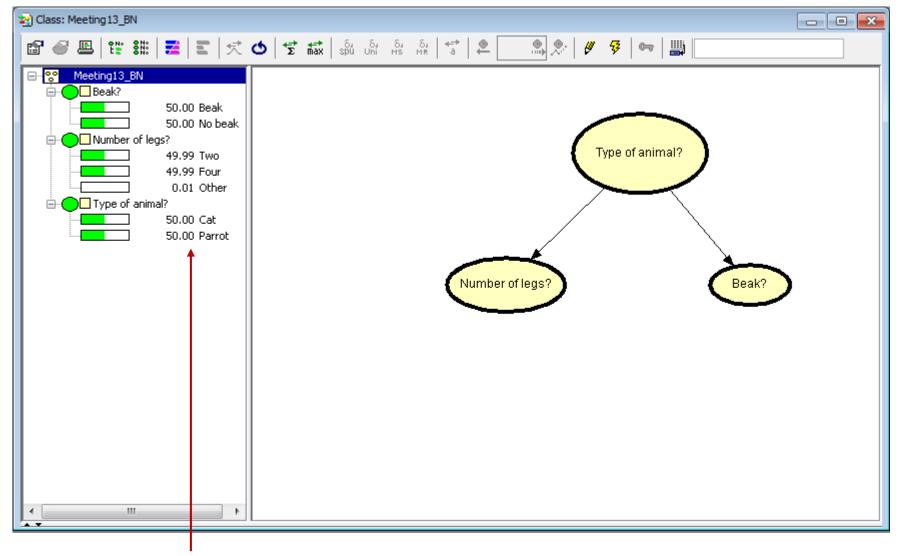
Files of type:

But it is also possible to use the more general NET format for compatibility with other software (like GeNIe) Now, the designed network should be "run". This means putting the probabilities set into "action"





Expand this list

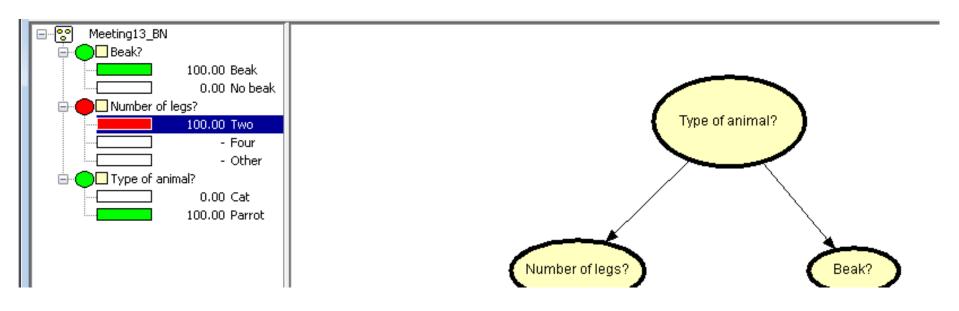


Here, we can read of the marginal probabilities (in %) of each state in each node. P(Beak|I), P(No Beak|I); P(Two legs|I), P(Four legs|I), P(Other|I) and (as previously assigned) P(Cat|I) and P(Parrot|I)

## Entering "evidence"

It is now possible to calculate updated (conditional) probabilities given a particular state in one or several of the nodes. This is called "to instantiate" a node to the state of interest and is done in the software by double-clicking on that state.

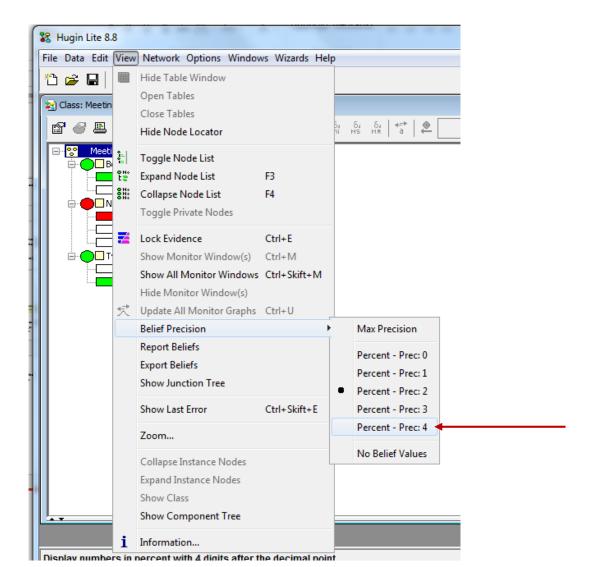
As an example, suppose that we obtain the information that the animal (Willie) has two legs. Then, we double-click on that state in the list.

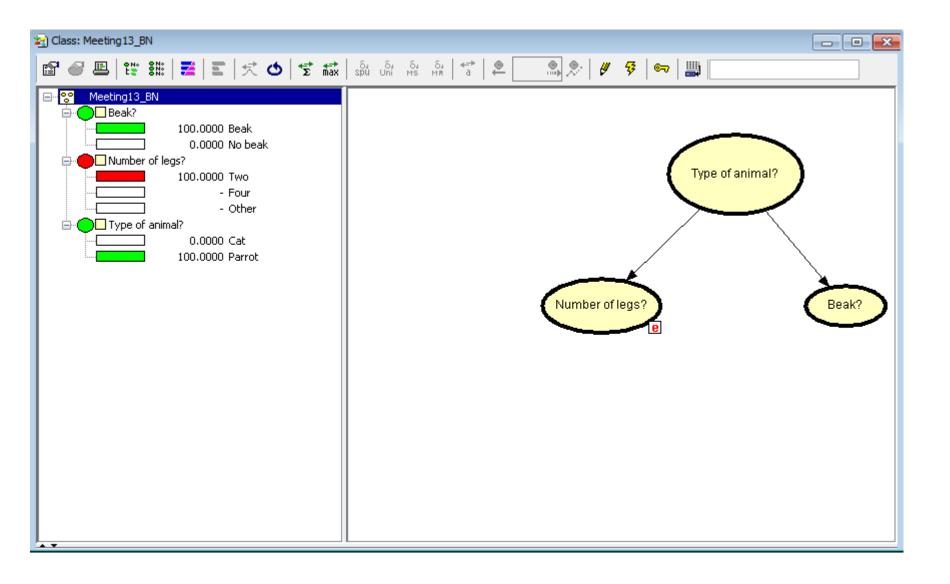


The bar colour of this state changes into read and its value to 100 %. The other bars also change values, i.e. the conditional probabilities given Willie has four legs. However, it seems they are either 100.00% or 0.00%. Could that be correct?

The precision used for displaying the numbers can be changed.

From the GUI menu select View, from the list select Belief Precision and in the following list select Percent – Prec: 4. *Other choice can of course also be made*.





Well, we have now the values 100.0000% and 0.0000% respectively. Trying Max Precision would not give more information.

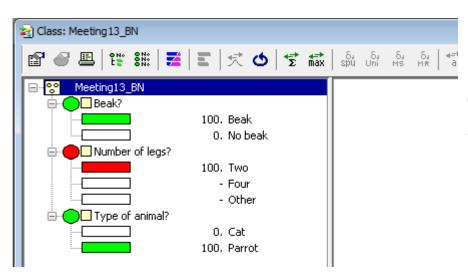
What probabilities are we computing here? Consider for instance the updated probability for  $A_2$ , i.e. "Willie is a parrot".

This probability is

$$P(A_2|B_1) = \frac{P(B_1|A_2) \cdot P(A_2)}{P(B_1|A_1) \cdot P(A_1) + P(B_1|A_2) \cdot P(A_2)} = \frac{0.9999 \cdot 0.5}{0 \cdot 0.5 + 0.9999 \cdot 0.5} = 1$$

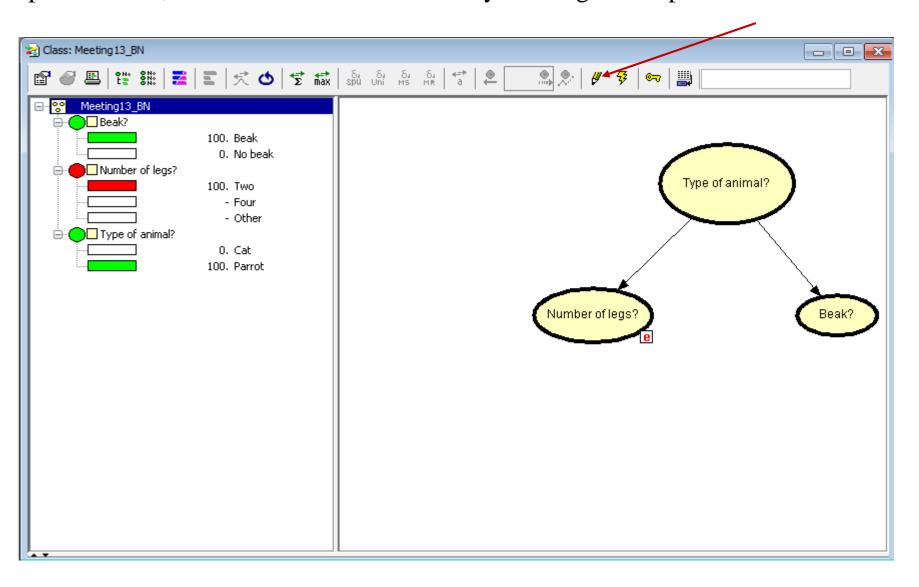
Hence, the probability is (maybe not so unexpected) exactly 100%

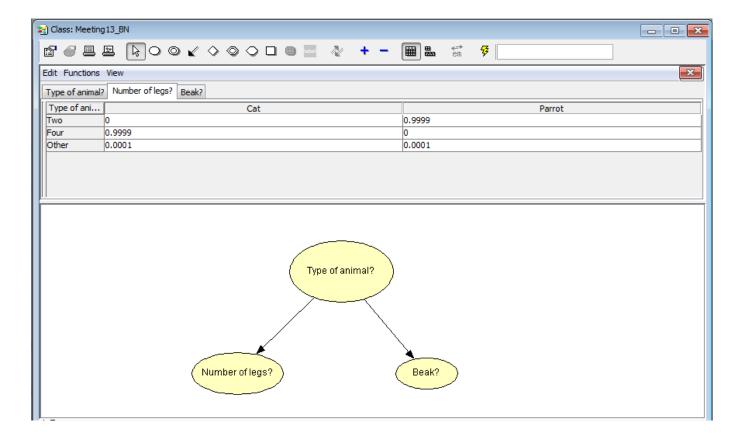
This can also be seen by setting Belief Precision to Max precision



The absence of displayed decimals indicate that the values displayed are exactly 100 and 0 respectively.

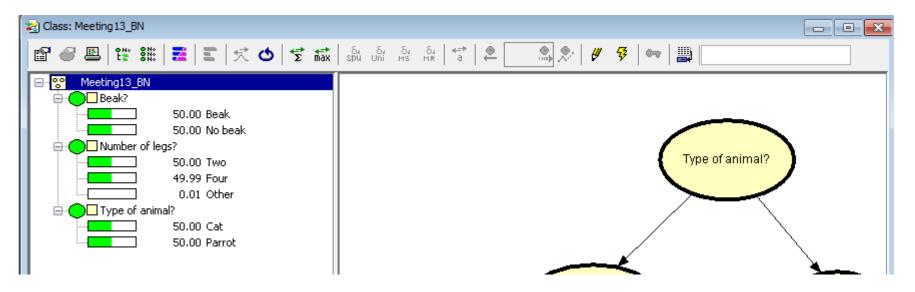
If we want to edit the network, e.g. adding nodes and/or changing assigned probabilities, we can return to Edit mode by clicking on the pencil icon.



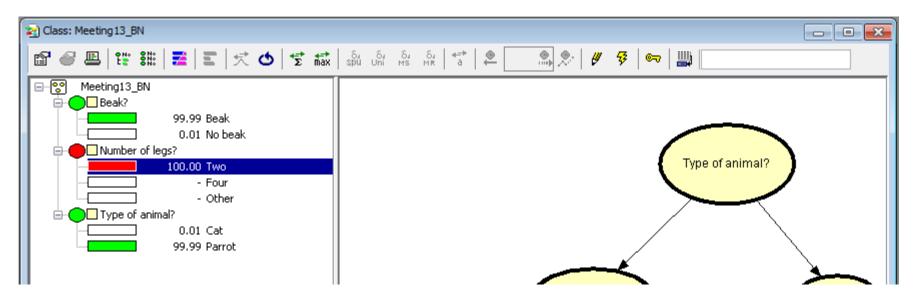


Now, change the conditional distribution of the number of legs given the type of animal is a cat to 0.0001, 0.9998 and 0.0001 respectively, and run network again.

Type of animal?	Number of legs?	Beak?
Type of ani	Cat	
Two	0.0	001
Four -	<del>0.9999</del> 0.9	998
Other	0.0001 0.0	001



Again, instantiate state "Two" in node Number of legs?



...and we can see that the updated probability for Willie being a parrot is now 0.9999.