

## Machine learning: lecture 20

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#### **Outline**

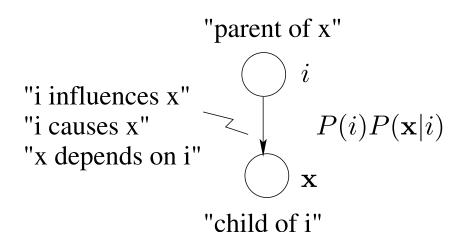
- Bayesian networks
  - examples, specification
  - graphs and independence
  - associated distribution



### Bayesian networks

 Bayesian networks are directed acyclic graphs, where the nodes represent variables and directed edges capture dependencies

A mixture model as a Bayesian network

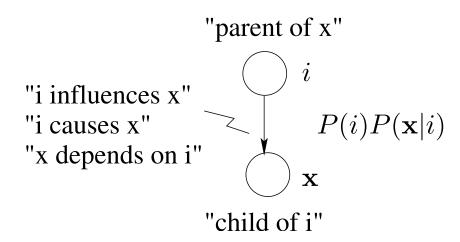




### Bayesian networks

 Bayesian networks are directed acyclic graphs, where the nodes represent variables and directed edges capture dependencies

A mixture model as a Bayesian network



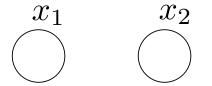
Graph semantics:

graph  $\Rightarrow$  separation properties  $\Rightarrow$  independence

Association with probability distributions:

independence  $\Rightarrow$  family of distributions

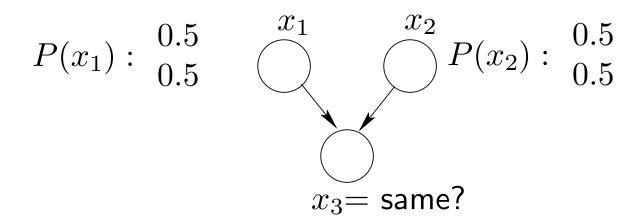




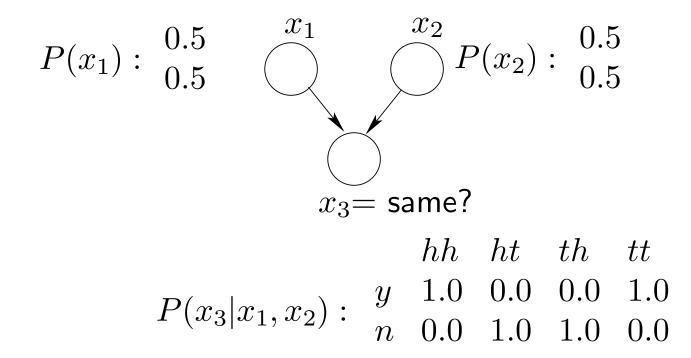


$$P(x_1): \begin{array}{ccc} 0.5 & x_1 & x_2 \\ 0.5 & \bigcirc & P(x_2): \begin{array}{c} 0.5 \\ 0.5 \end{array}$$

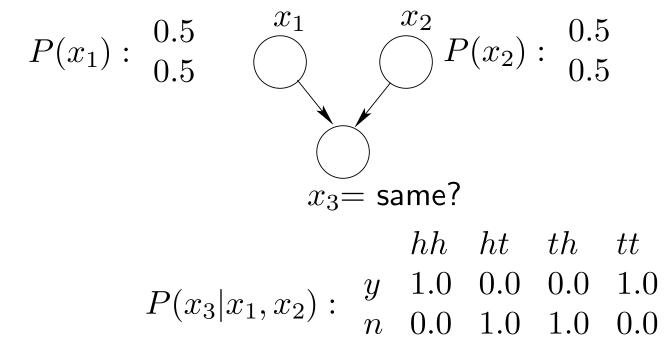










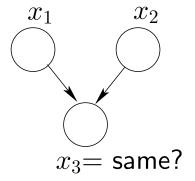


- Two levels of description
  - 1. graph structure (dependencies, independencies)
  - 2. associated probability distribution



## **Example cont'd**

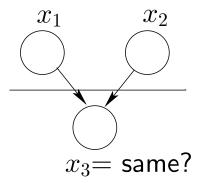
• What can the graph alone tell us?





## Example cont'd

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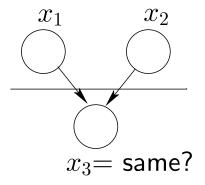


•  $x_1$  and  $x_2$  are marginally independent

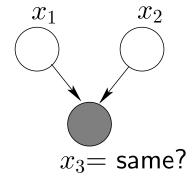


#### Example cont'd

What can the graph alone tell us?



•  $x_1$  and  $x_2$  are marginally independent



•  $x_1$  and  $x_2$  become *dependent* if we know  $x_3$  (the dependence concerns our beliefs about the outcomes)

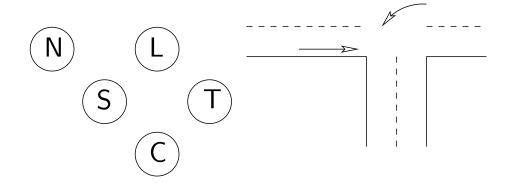


N = X is nice?

L = traffic light

S = X decides to stop?

T =the other car turns left?



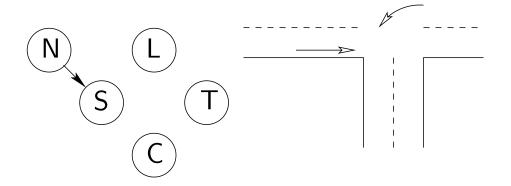


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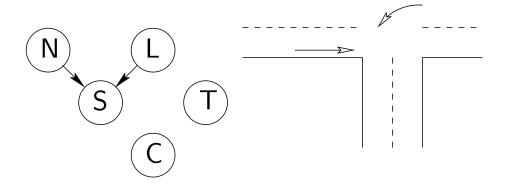


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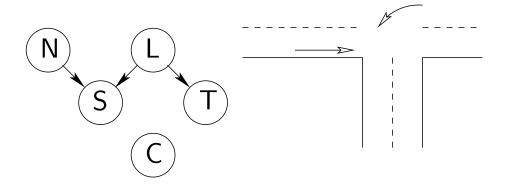


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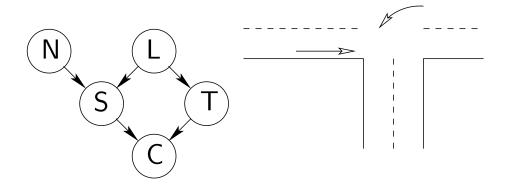


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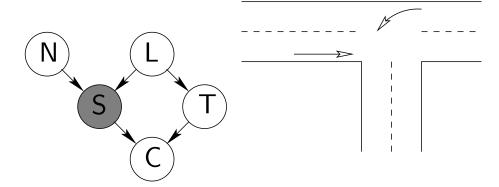
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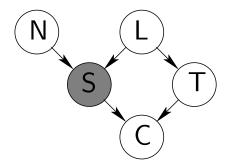
C = crash?



 If we only know that X decided to stop, can X's character (variable N) tell us anything about the other car turning (variable T)?

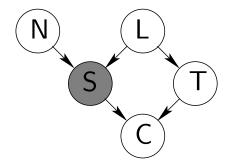


ullet Are N and T independent given S?



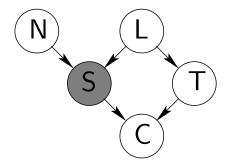


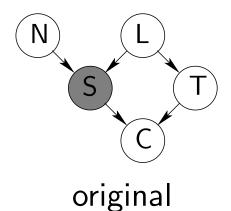
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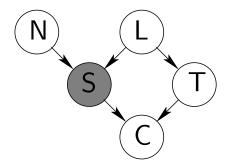
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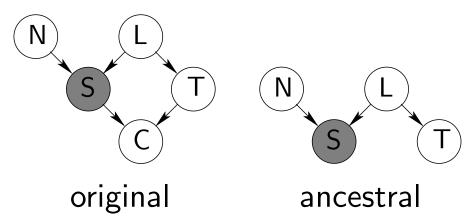






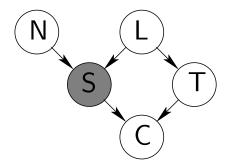
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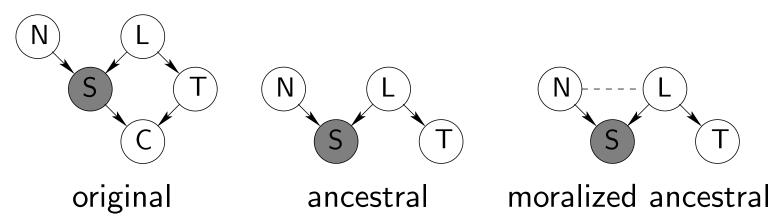






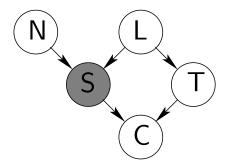
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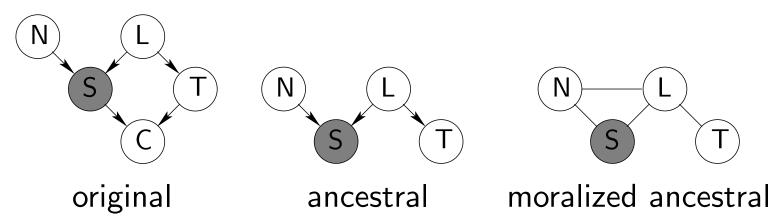






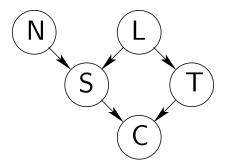
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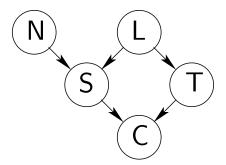
## **Graphs and distributions**



 A graph is a compact representation of a large collection of independence properties



#### **Graphs and distributions**



 A graph is a compact representation of a large collection of independence properties

**Theorem:** Any probability distribution that is consistent with a directed graph G has to factor according to "node given parents":

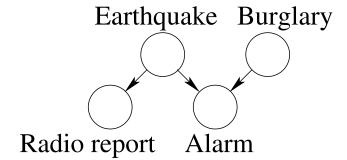
$$P(\mathbf{x}|G) = \prod_{i=1}^{d} P(x_i|\mathbf{x}_{pa_i})$$

where  $\mathbf{x}_{pa_i}$  are the *parents* of  $x_i$  and d is the number of nodes (variables) in the graph.



# **Explaining away phenomenon**

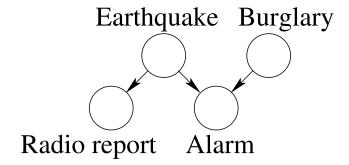
Model



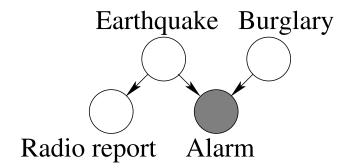


## **Explaining away phenomenon**

Model



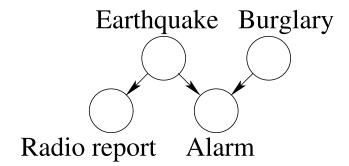
• Evidence, competing causes



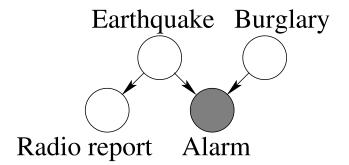


## **Explaining away phenomenon**

Model



Evidence, competing causes



Additional evidence and explaining away

