

# **ML 2016/17**

## **Exercise 6:**

# **Bayesian Networks**



05/12/16

# General information

- The assignments are not graded on a scale: it's simply pass/no pass
  - If one homework is not sufficient you can simply redo it
- All assignments must be delivered one month before **you** take the exam
- Submission through email: send to [fabiom.carlucci@dis.uniroma1.it](mailto:fabiom.carlucci@dis.uniroma1.it)
- Questions can be written to same email address.
- Office hours to meet in person: **Wednesday** at B004 (Via Ariosto, the door in front of library), 10AM-12PM.
- *There is no need to replicate exactly the images I show!*

# HW6: BN

For a change, no *sklearn* today!

<http://www.aispace.org/bayes/version5.1.9/bayes.jar>

To launch, double click (on Windows) or type in console:

```
java -jar bayes.jar
```

Once you complete the experience, send the report to [fabiom.carlucci@dis.uniroma1.it](mailto:fabiom.carlucci@dis.uniroma1.it) with subject: “[ML1617] BN report”

# Bayesian Networks in practice

- We will deal only with Bayesian Networks
- They help with
  - Factorization of the probability distribution
  - Visualization of Conditional independence properties
  - inference algorithms (ex: Markov Chain Montecarlo)

Conditional independence allows us to deal with models which would otherwise be intractable

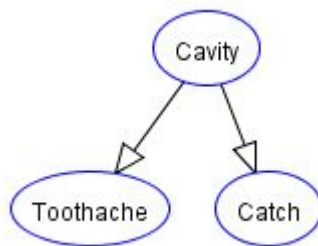
# From “AI: A Modern Approach”

A Bayesian network is a directed graph in which each node is annotated with quantitative probability information. The full specification is as follows:

1. Each node corresponds to a random variable, which may be discrete or continuous.
2. A set of directed links or arrows connects pairs of nodes. If there is an arrow from node  $X$  to node  $Y$ ,  $X$  is said to be a parent of  $Y$ . The graph has no directed cycles (and hence is a directed acyclic graph, or DAG).
3. Each node  $X_i$  has a conditional probability distribution  $P(X_i \mid \text{Parents}(X_i))$  that quantifies the effect of the parents on the node

# BN in practice

- Intuitively causes are parents of effects - X has direct influence on Y
- Catch given Cavity is conditionally independent from Toothache
- $P(\text{toothache}, \text{cavity}, \text{catch}) = P(\text{cavity})P(\text{toothache}|\text{cavity})P(\text{catch}|\text{cavity})$



# The tool - part 1

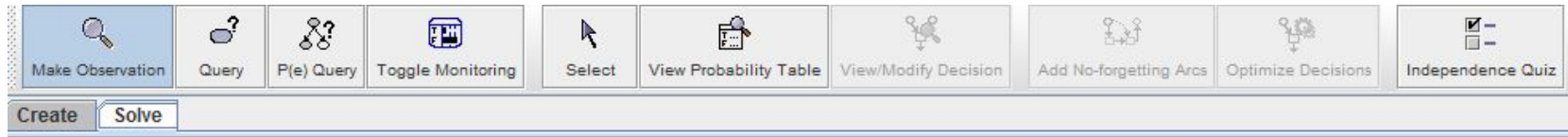


The tool's website: <http://www.aispace.org/bayes/>

It allows you to

- Create nodes
- Create arcs
- Set properties
- Set Probability tables
- *(continues...)*

# The tool - part 2



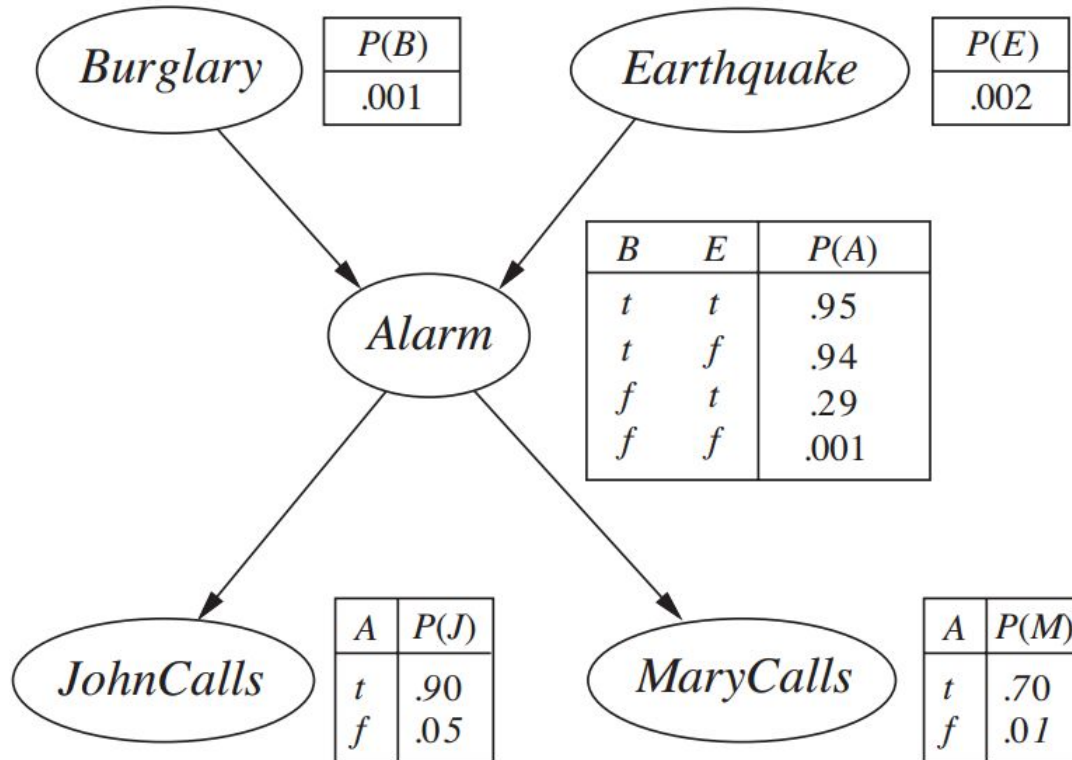
- Make observations
- View probabilities
- Quiz yourself
- And other



# Burglars and earthquakes example

- There is a probability of 0.001 that there will be a burglary
- There is a probability of 0.002 that there will be an earthquake
- There is an alarm, that will activate with probability
  - 0.95 if there is a burglary *and* an earthquake
  - 0.94 if there is only a burglary
  - 0.29 if there is only an earthquake
  - 0.001 if no earthquake and no burglary
- John will call you with probability 0.9 if the alarm is sounding. There is a 0.05 chance that he will call if even if no alarm has sounded (he imagines things...)
- Mary is a little deaf and will call you with probability 0.7 if the alarms sounds. If the alarms doesn't sounds there is still a small chance that she will call you by mistake ( probability 0.01)

# Burglars and earthquakes example



# Burglars and earthquakes Joint Probability

By the chain rule:

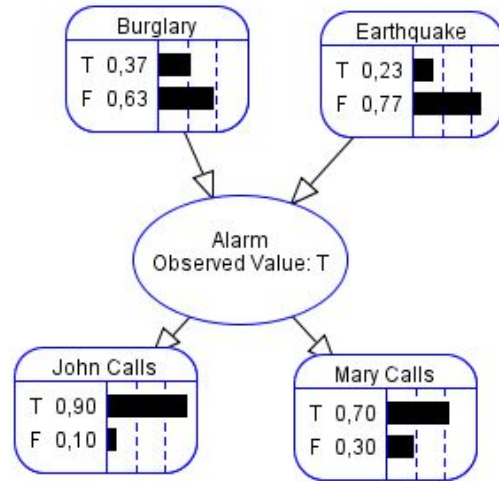
$$P(b, e, a, j, m) = P(m|b, e, a, j)P(b|e, a, j)P(e|a, j)P(a|j)P(j)$$

exploiting the conditional independence defined in the BN:

$$P(b, e, a, j, m) = P(b)P(e)P(a|b, e)P(j|a)P(m|a)$$

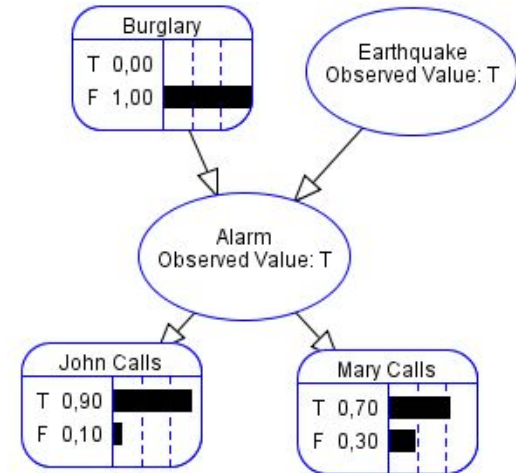
# We can make observations

What if we hear the alarm?



What if we also hear an earthquake?

The probability of Burglary changes, but John Calls doesn't, why?



# Assignment 1

1. Draw the Bayesian network corresponding to the following:
  - a. There is a 0.5 probability that it is cloudy
  - b. If it's cloudy,  $P(\text{Rain}) = 0.8$  (The probability that it rains is 0.8)
  - c. If it's not cloudy, then  $P(\text{Rain}) = 0.1$
  - d. If it's cloudy, the probability that the sprinklers starts (  $P(\text{Sprinklers})$  ) is only 0.4
  - e. If it's not cloudy, then  $P(\text{Sprinklers}) = 0.9$
  - f. If it rains and the sprinklers are activated, then  $P(\text{WetGrass}) = 0.99$
  - g. If it rains, but the sprinklers are off, then  $P(\text{WetGrass}) = 0.9$
  - h. If it doesn't rain, but the sprinklers work, then  $P(\text{WetGrass}) = 0.9$
  - i. If it neither rains, nor the sprinkler works, then  $P(\text{WetGrass}) = 0$
2. Write the joint probability function for this graph
3. If  $\text{WetGrass} = \text{True}$  and  $\text{Cloudy} = \text{False}$ , what is the probability it rained? The probability the sprinklers were on?

# Assignment 2

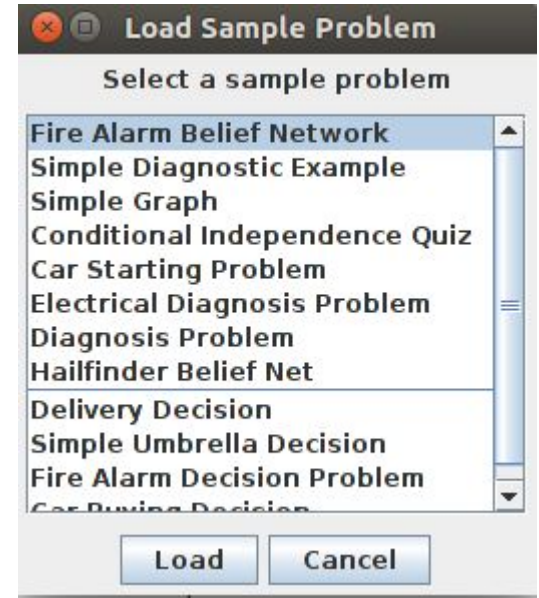
1. Draw the Bayesian network corresponding to the following:

We have a bag of three biased coins a, b, and c with probabilities of coming up heads of 20%, 60%, and 80%, respectively. One coin is drawn randomly from the bag (with equal likelihood of drawing each of the three coins), and then the coin is flipped three times to generate the outcomes  $X_1$ ,  $X_2$ , and  $X_3$ .

2. Calculate which coin was most likely to have been drawn from the bag if the observed flips come out heads twice and tails once [note: you should use the tool]

# Assignment 3

1. Load the sample problem “Fire Alarm Belief Network”
2. Write down the joint probability
3. Change the network by adding a node representing the probability that someone will call their mother if the alarm goes off



**Your turn now! Questions?**

