

Reasoning under Uncertainty

The intelligent way to handle the unknown

COURSE: CS60045

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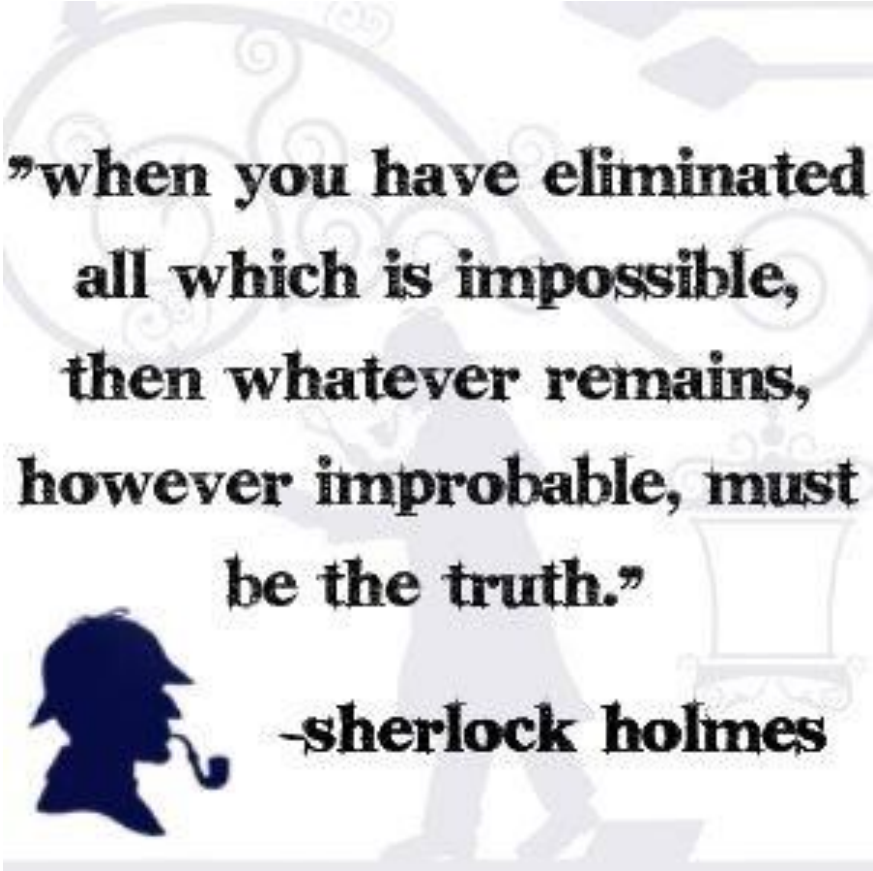
Logical Deduction versus Induction

DEDUCTION

- Commonly associated with *formal logic*
- Involves reasoning from known premises to a conclusion
- The conclusions reached are inevitable, certain, inescapable

INDUCTION

- Commonly known as *informal logic* or *everyday argument*
- Involves drawing uncertain inferences based on probabilistic reasoning
- The conclusions reached are probable, reasonable, plausible, believable

A faint, light-colored silhouette of Sherlock Holmes in his iconic deerstalker hat and smoking a pipe, standing in a room with a chair and a table. The silhouette is positioned behind the text.

**“when you have eliminated
all which is impossible,
then whatever remains,
however improbable, must
be the truth.”**

-sherlock holmes

Handling uncertain knowledge

- **Classical first order logic has no room for uncertainty**

$\forall p \text{ Symptom}(p, \text{Toothache}) \Rightarrow \text{Disease}(p, \text{Cavity})$

- **Not correct – toothache can be caused in many other cases**
- **In first order logic we have to include all possible causes**

**$\forall p \text{ Symptom}(p, \text{Toothache}) \Rightarrow \text{Disease}(p, \text{Cavity}) \vee \text{Disease}(p, \text{GumDisease})$
 $\vee \text{Disease}(p, \text{ImpactedWisdom}) \vee \dots$**

- **Similarly, Cavity does not always cause Toothache, so the following is also not true**

$\forall p \text{ Disease}(p, \text{Cavity}) \Rightarrow \text{Symptom}(p, \text{Toothache})$

Reasons for using probability

- **Specification becomes too large**
 - It is too much work to list the complete set of antecedents or consequents needed to ensure an exception-less rule
- **Theoretical ignorance**
 - The complete set of antecedents is not known
- **Practical ignorance**
 - The truth of the antecedents is not known, but we still wish to reason

Predicting versus Diagnosing

- Probabilistic reasoning can be used for predicting outcomes (*from cause to effect*)
 - Given that I have a cavity, what is the chance that I will have toothache?
- Probabilistic reasoning can also be used for diagnosis (*from effect to cause*)
 - Given that I am having toothache, what is the chance that it is being caused by a cavity?

We need a methodology for reasoning that can work both ways.

Axioms of Probability

1. All probabilities are between 0 and 1: $0 \leq P(A) \leq 1$
2. $P(\text{True}) = 1$ and $P(\text{False}) = 0$
3. $P(A \vee B) = P(A) + P(B) - P(A \wedge B)$

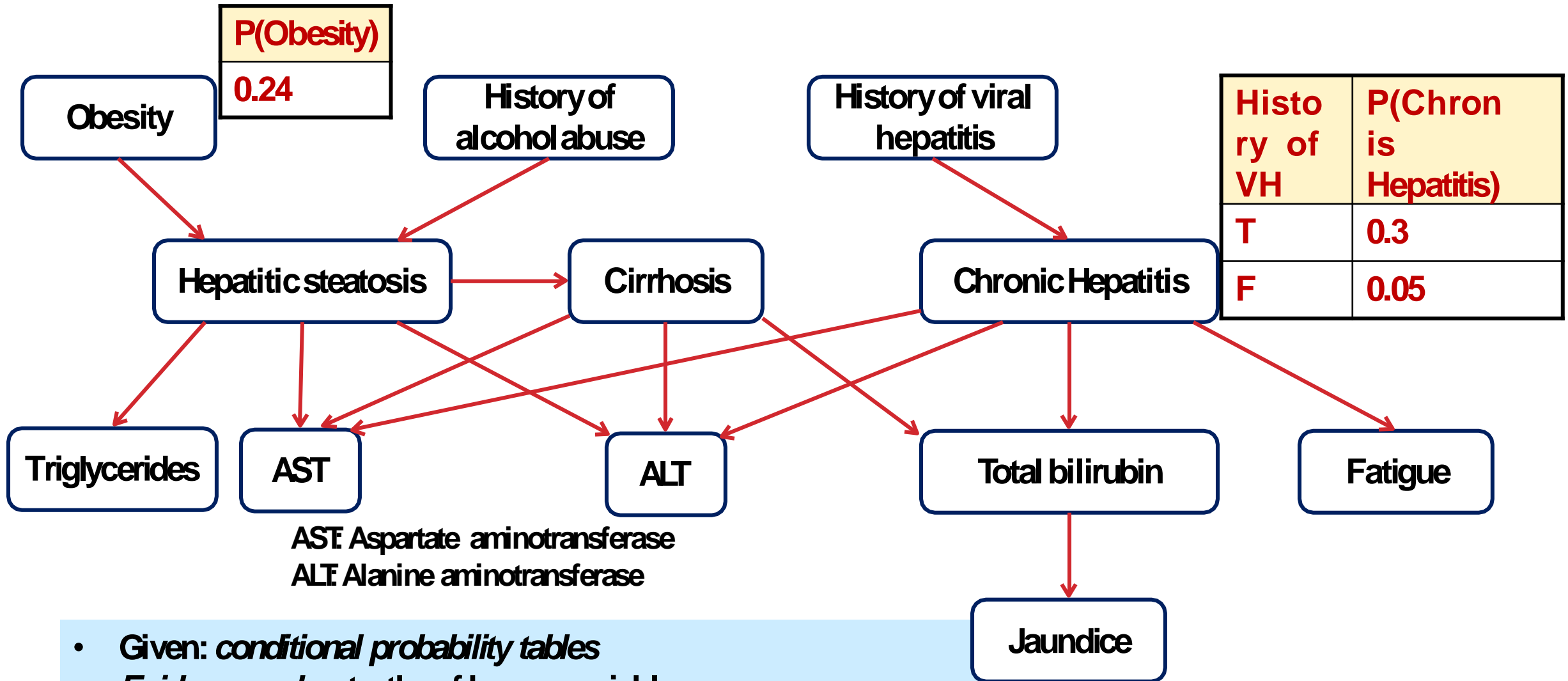
Bayes' Rule

$$P(A \wedge B) = P(A | B) P(B)$$

$$P(A \wedge B) = P(B | A) P(A)$$

$$P(B | A) = \frac{P(A | B) P(B)}{P(A)}$$

Bayesian Belief Network



- Given: *conditional probability tables*
- Evidence nodes: truths of known variables
- Goal: *Find probabilities of other variables and/or their combinations*

Belief Networks

A belief network is a graph with the following:

- **Nodes:** Set of random variables
- **Directed links:** The intuitive meaning of a link from node X to node Y is that X has a direct influence on Y

Each node has a **conditional probability table** that quantifies the effects that the parent have on the node.

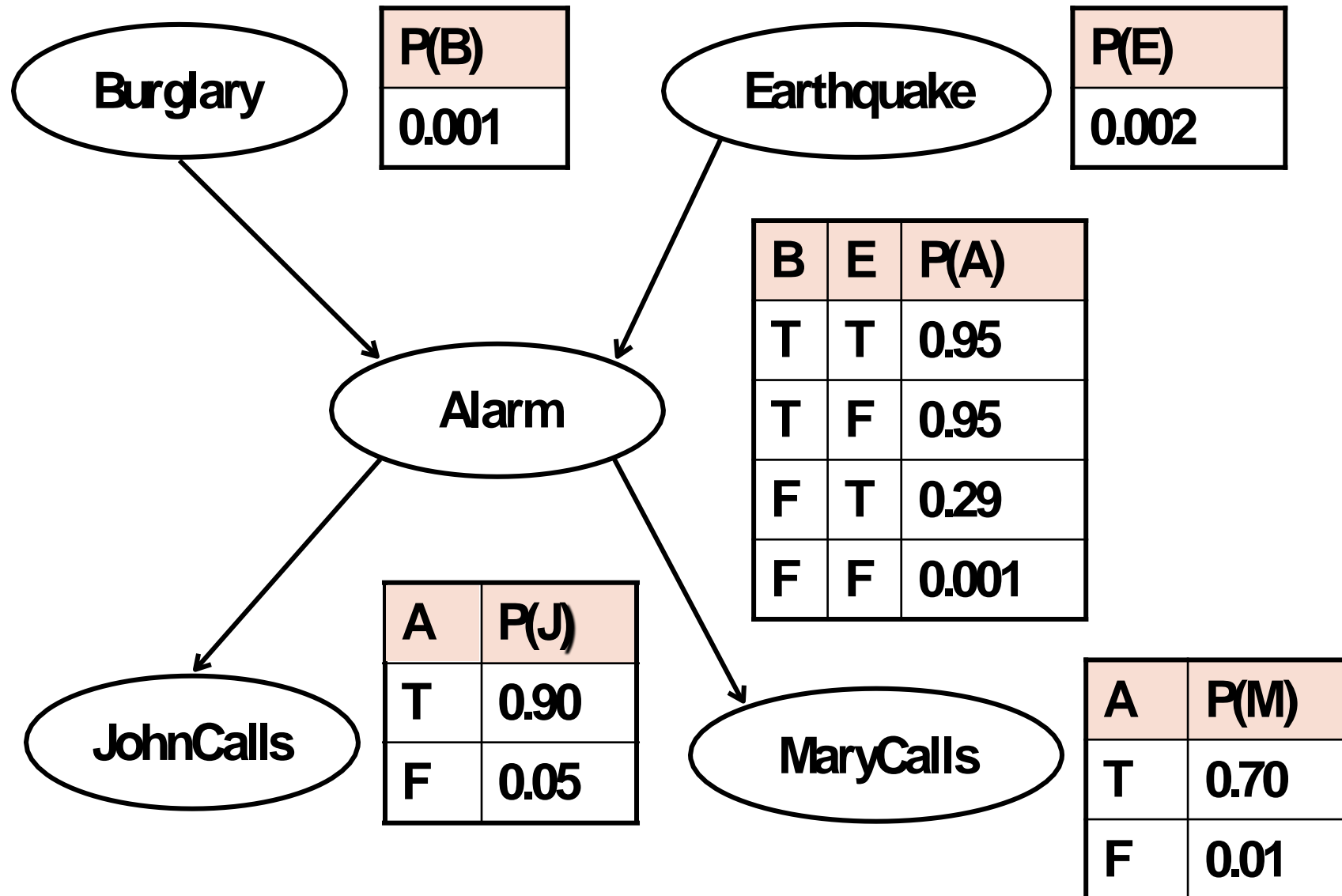
The graph has no directed cycles. It is a *directed acyclic graph* (DAG).

Classical Example

- **Burglar alarm at home**
 - Fairly reliable at detecting a burglary
 - Responds at times to minor earthquakes
- **Two neighbors, on hearing alarm, calls police**
 - John always calls when he hears the alarm, but sometimes confuses the telephone ringing with the alarm and calls then, too.
 - Mary likes loud music and sometimes misses the alarm altogether



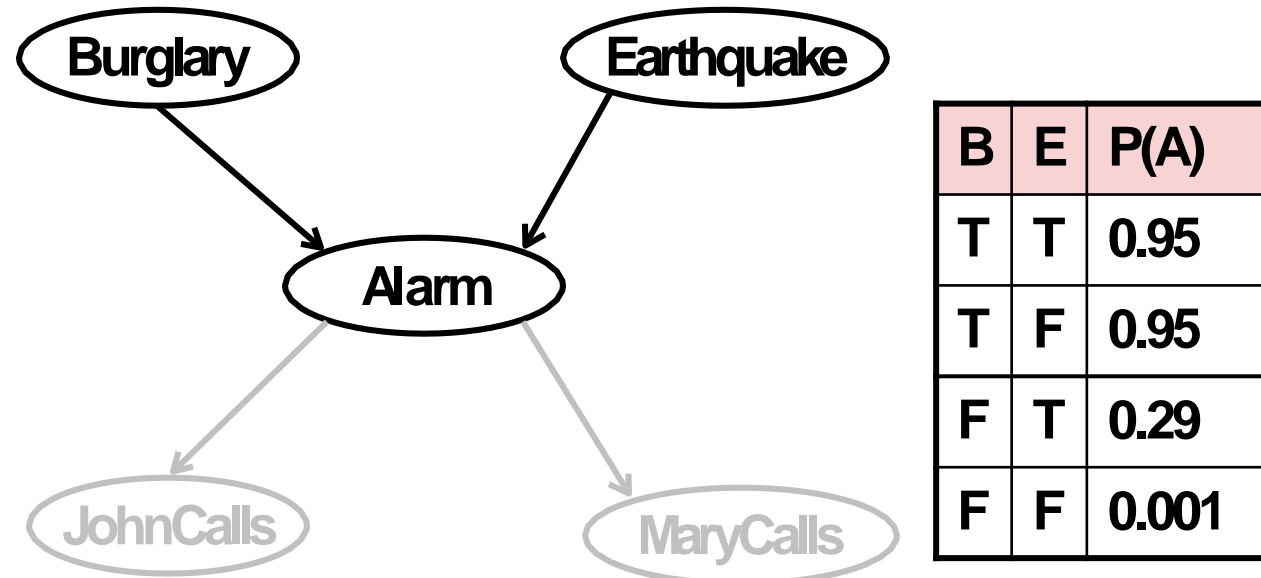
Belief Network Example



The joint probability distribution

- A generic entry in the joint probability distribution $P(x_1, \dots, x_n)$ is given by:

$$P(x_1, \dots, x_n) = \prod_{i=1}^n P(x_i \mid \text{Parents}(X_i))$$



The joint probability distribution

- Probability of the event that the alarm has sounded but neither a burglary nor an earthquake has occurred, and both Mary and John call:

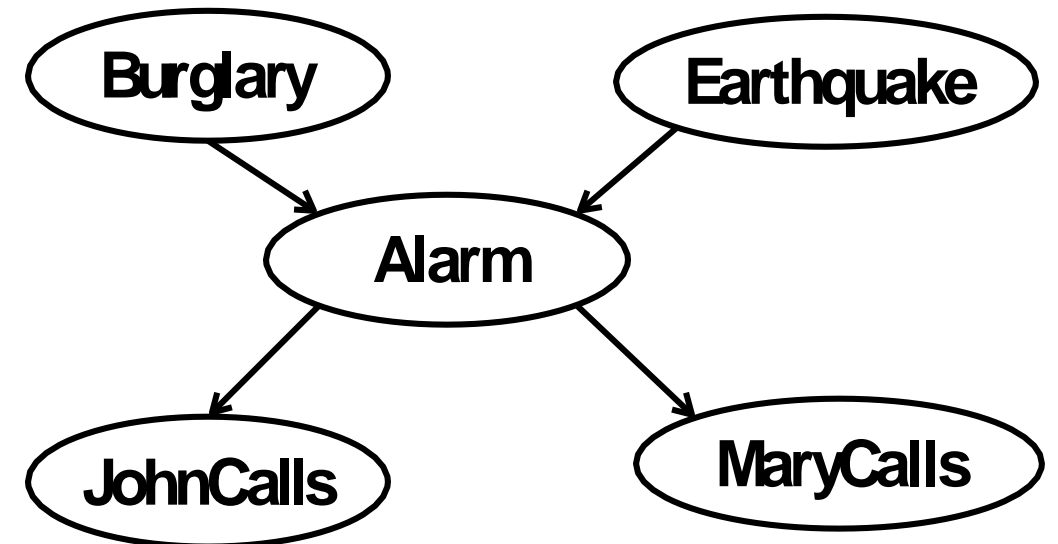
$$\begin{aligned} &P(J \wedge M \wedge A \wedge \neg B \wedge \neg E) \\ &= P(J | A) P(M | A) P(A | \neg B \wedge \neg E) P(\neg B) P(\neg E) \\ &= 0.9 \times 0.7 \times 0.001 \times 0.999 \times 0.998 \\ &= 0.00062 \end{aligned}$$

B	E	P(A)
T	T	0.95
T	F	0.95
F	T	0.29
F	F	0.001

A	P(J)
T	0.90
F	0.05

A	P(M)
T	0.70
F	0.01

P(E)	P(B)
0.002	0.001



The joint probability distribution

- Computation of the probabilities of several different event combinations of the Burglary-Alarm belief network example:

$$P(B) = 0.001$$

$$P(B') = 1 - P(B) = 0.999$$

$$P(E) = 0.002$$

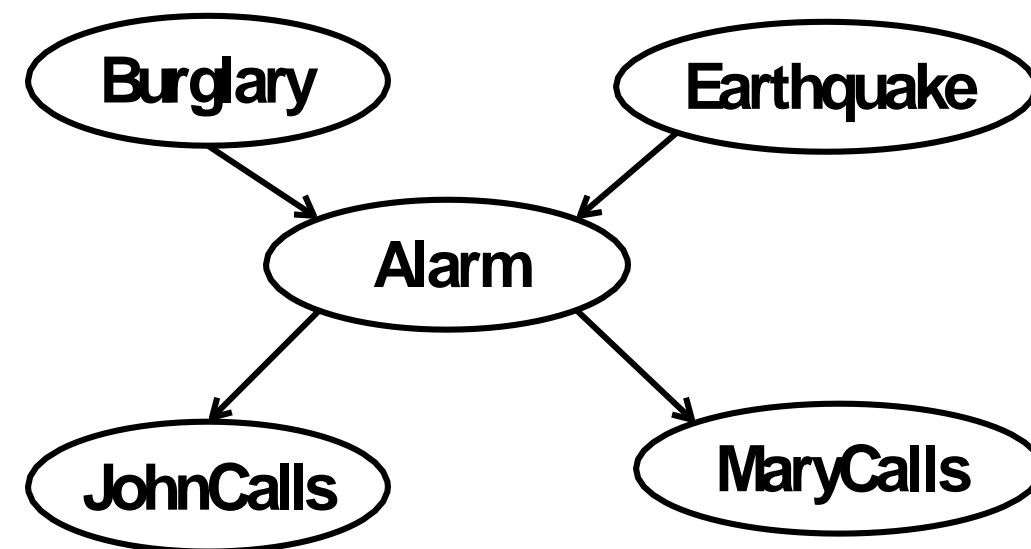
$$P(E') = 1 - P(E) = 0.998$$

B	E	P(A)
T	T	0.95
T	F	0.95
F	T	0.29
F	F	0.001

A	P(J)
T	0.90
F	0.05

A	P(M)
T	0.70
F	0.01

P(E)	P(B)
0.002	0.001



The joint probability distribution

- Computation of the probabilities of several different event combinations of the Burglary-Alarm belief network example:

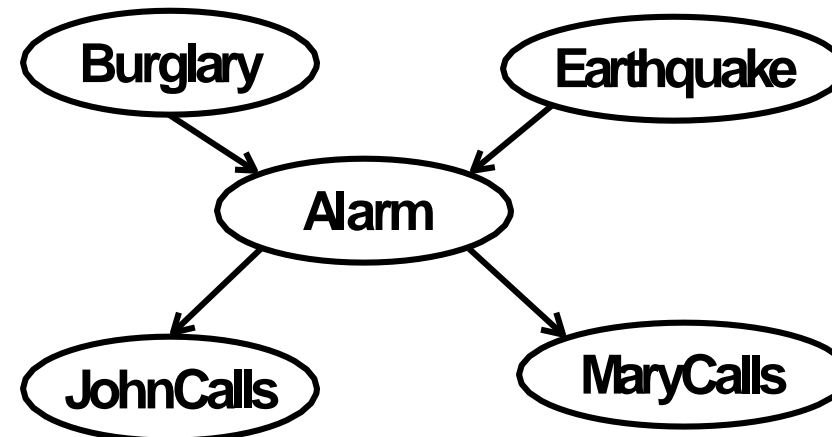
$$\begin{aligned}
 P(A) &= P(AB'E') + P(AB'E) + P(ABE') + P(ABE) \\
 &= P(A | B'E').P(B'E') + P(A | B'E).P(B'E) + P(A | BE').P(BE') + P(A | BE).P(BE) \\
 &= 0.001 \times 0.999 \times 0.998 + 0.29 \times 0.999 \times 0.002 + 0.95 \times 0.001 \times 0.998 + 0.95 \times 0.001 \times 0.002 \\
 &= 0.001 + 0.0006 + 0.0009 = 0.0025
 \end{aligned}$$

B	E	P(A)
T	T	0.95
T	F	0.95
F	T	0.29
F	F	0.001

A	P(J)
T	0.90
F	0.05

A	P(M)
T	0.70
F	0.01

P(E)	P(B)
0.002	0.001



The joint probability distribution: *Find P(J)*

$$\begin{aligned}
 P(J) &= P(JA) + P(JA') \\
 &= P(J | A).P(A) + P(J | A').P(A') \\
 &= 0.9 \times 0.0025 + 0.05 \times (1 - 0.0025) \\
 &= 0.052125
 \end{aligned}$$

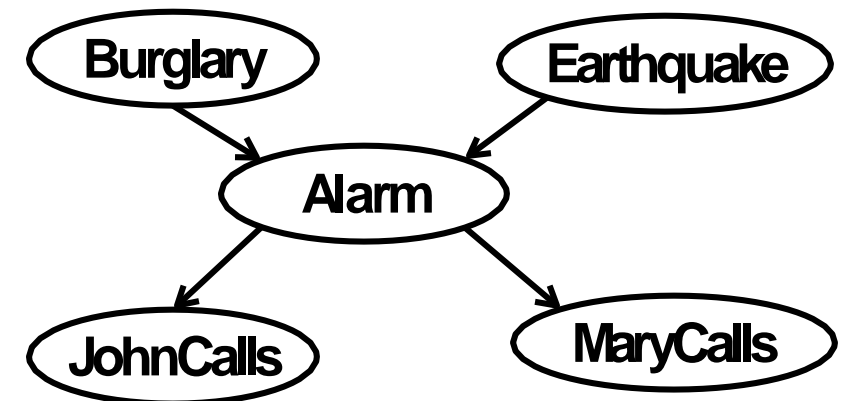
$$\begin{aligned}
 P(AB) &= P(ABE) + P(ABE') = 0.95 \times 0.001 \times 0.002 + 0.95 \times 0.001 \times 0.998 \\
 &= 0.00095
 \end{aligned}$$

B	E	P(A)
T	T	0.95
T	F	0.95
F	T	0.29
F	F	0.001

A	P(J)
T	0.90
F	0.05

A	P(M)
T	0.70
F	0.01

P(E)	P(B)
0.002	0.001



The joint probability distribution: Find $P(A'B)$ and $P(AE)$

$$\begin{aligned}
 P(A'B) &= P(A'BE) + P(A'BE') \\
 &= P(A' | BE) \cdot P(BE) + P(A' | BE') \cdot P(BE') \\
 &= (1 - 0.95) \times 0.001 \times 0.002 \\
 &\quad + (1 - 0.95) \times 0.001 \times 0.998 \\
 &= 0.00005
 \end{aligned}$$

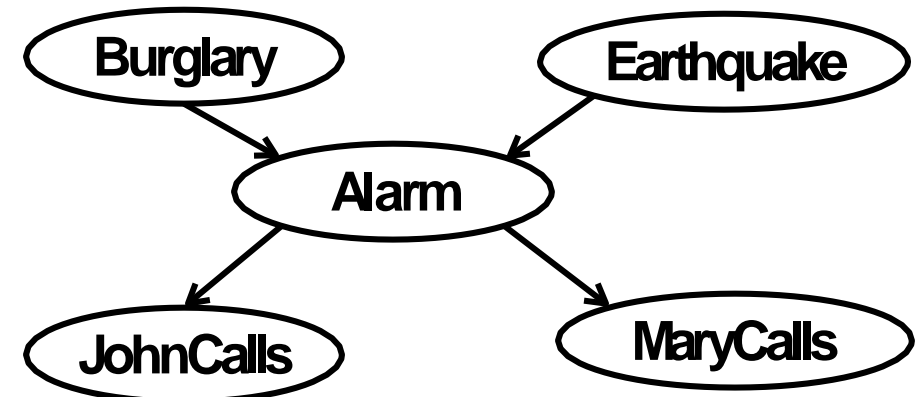
$$\begin{aligned}
 P(AE) &= P(AEB) + P(AEB') \\
 &= 0.95 \times 0.001 \times 0.002 + 0.29 \times 0.999 \times 0.002 = 0.00058
 \end{aligned}$$

B	E	P(A)
T	T	0.95
T	F	0.95
F	T	0.29
F	F	0.001

A	P(J)
T	0.90
F	0.05

A	P(M)
T	0.70
F	0.01

P(E)	P(B)
0.002	0.001



The joint probability distribution

$$\begin{aligned}
 P(AE') &= P(AE'B) + P(AE'B') \\
 &= 0.95 \times 0.001 \times 0.998 + 0.001 \times 0.999 \times 0.998 \\
 &= 0.001945
 \end{aligned}$$

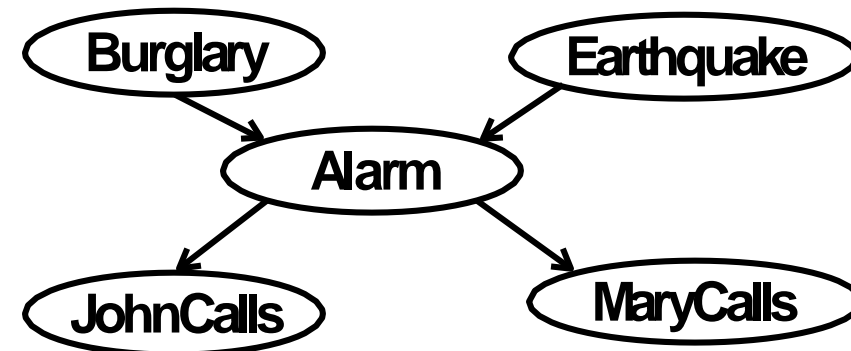
$$\begin{aligned}
 P(A'E') &= P(A'E'B) + P(A'E'B') \\
 &= P(A' | BE') \cdot P(BE') + P(A' | B'E') \cdot P(B'E') \\
 &= (1 - 0.95) \times 0.001 \times 0.998 + (1 - 0.001) \times 0.999 \times 0.998 = 0.996
 \end{aligned}$$

B	E	P(A)
T	T	0.95
T	F	0.95
F	T	0.29
F	F	0.001

A	P(J)
T	0.90
F	0.05

A	P(M)
T	0.70
F	0.01

P(E)	P(B)
0.002	0.001



The joint probability distribution: *Find* P(JB)

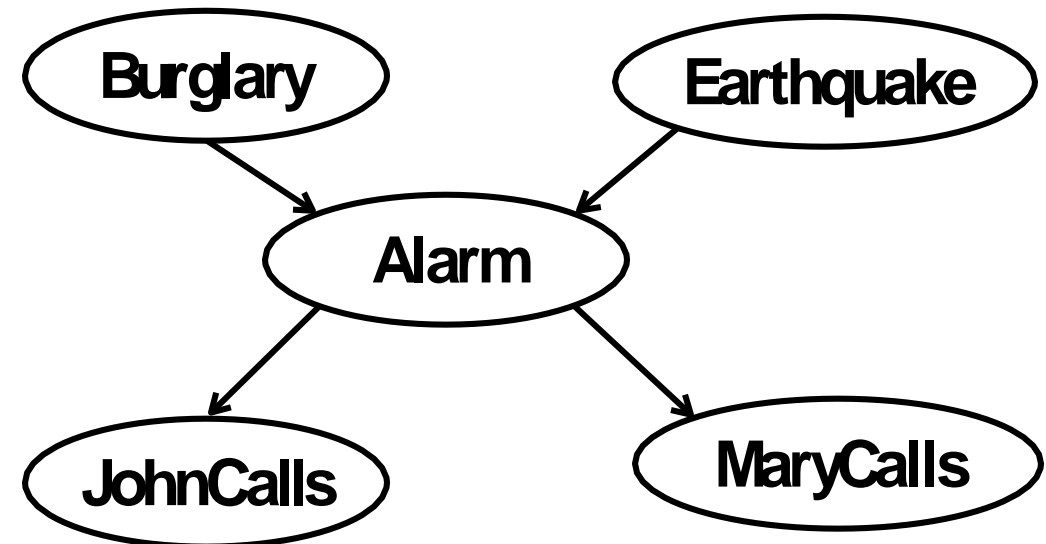
$$\begin{aligned} P(JB) &= P(JBA) + P(JBA') \\ &= P(J | AB).P(AB) + P(J | A'B).P(A'B) \\ &= P(J | A).P(AB) + P(J | A').P(A'B) \\ &= 0.9 \times 0.00095 + 0.05 \times 0.00005 \\ &= 0.00086 \end{aligned}$$

B	E	P(A)
T	T	0.95
T	F	0.95
F	T	0.29
F	F	0.001

A	P(J)
T	0.90
F	0.05

A	P(M)
T	0.70
F	0.01

P(E)	P(B)
0.002	0.001



The joint probability distribution

- Computation of the probabilities of several different event combinations of the Burglary-Alarm belief network example:

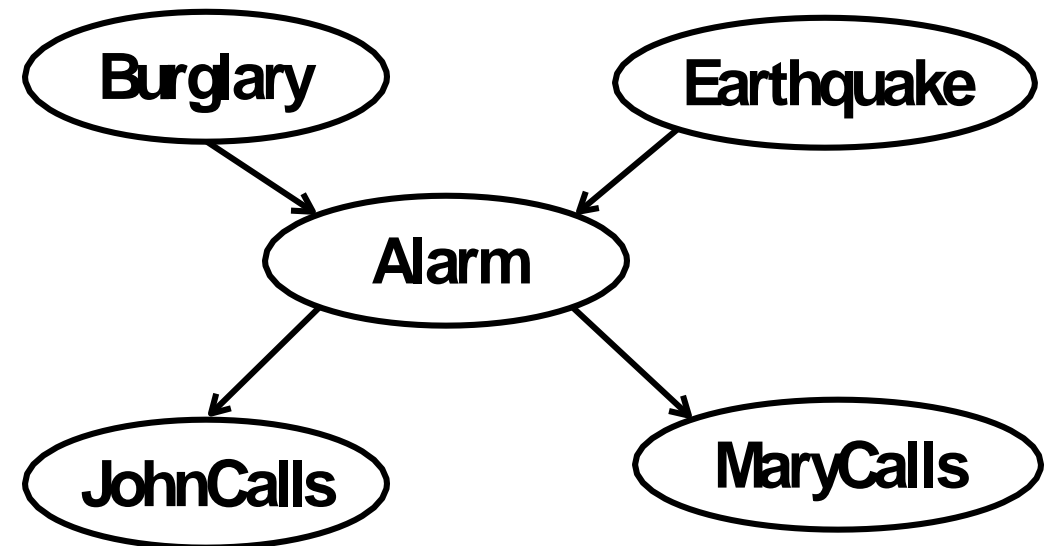
$$P(J | B) = P(JB) / P(B) = 0.00086 / 0.001 = 0.86$$

B	E	P(A)
T	T	0.95
T	F	0.95
F	T	0.29
F	F	0.001

A	P(J)
T	0.90
F	0.05

A	P(M)
T	0.70
F	0.01

P(E)	P(B)
0.002	0.001



The joint probability distribution

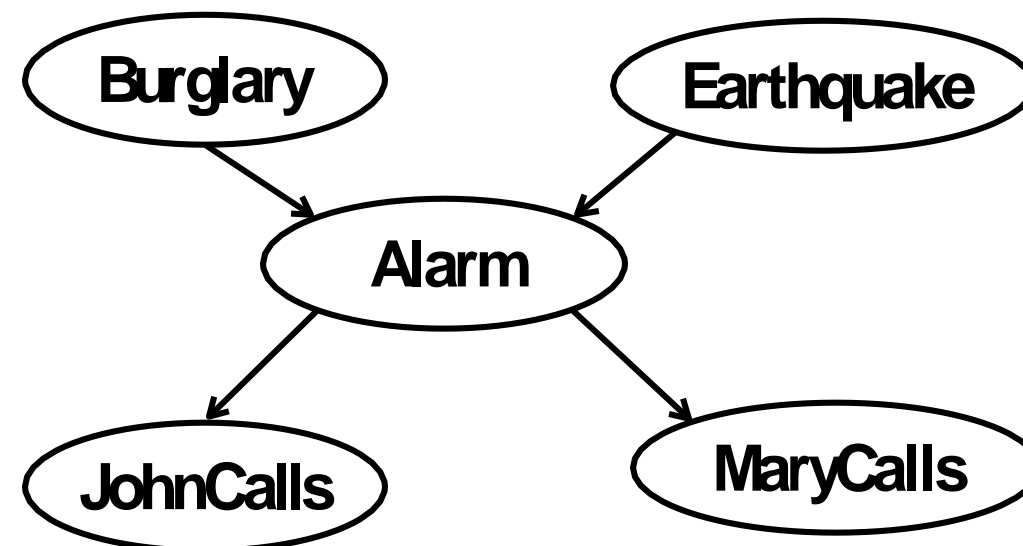
$$\begin{aligned}P(MB) &= P(MBA) + P(MBA') \\&= P(M | AB).P(AB) + P(M | A'B).P(A'B) \\&= P(M | A).P(AB) + P(M | A').P(A'B) \\&= 0.7 \times 0.00095 + 0.01 \times 0.00005 \\&= 0.00067\end{aligned}$$

B	E	P(A)
T	T	0.95
T	F	0.95
F	T	0.29
F	F	0.001

A	P(J)
T	0.90
F	0.05

A	P(M)
T	0.70
F	0.01

P(E)	P(B)
0.002	0.001



The joint probability distribution

$$P(M | B) = P(MB) / P(B) = 0.00067 / 0.001 = 0.67$$

$$P(B | J) = P(JB) / P(J) = 0.00086 / 0.052125 = 0.016$$

$$P(B | A) = P(AB) / P(A) = 0.00095 / 0.0025 = 0.38$$

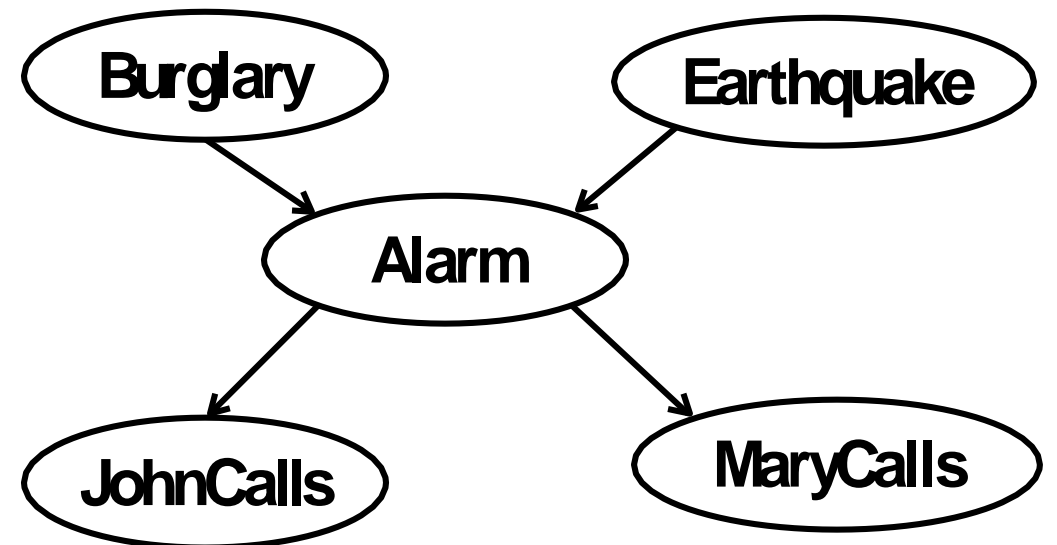
$$\begin{aligned} P(B | AE) &= P(ABE) / P(AE) = [P(A | BE) \cdot P(BE)] / P(AE) \\ &= [0.95 \times 0.001 \times 0.002] / 0.00058 \\ &= 0.003 \end{aligned}$$

B	E	P(A)
T	T	0.95
T	F	0.95
F	T	0.29
F	F	0.001

A	P(J)
T	0.90
F	0.05

A	P(M)
T	0.70
F	0.01

P(E)	P(B)
0.002	0.001



The joint probability distribution

- Computation of the probabilities of several different event combinations of the Burglary-Alarm belief network example:

$$P(AJE') = P(J | AE').P(AE') = P(J | A).P(AE')$$

$$= 0.9 \times 0.001945 = 0.00175$$

$$P(A'JE') = P(J | A'E').P(A'E') = P(J | A').P(A'E')$$

$$= 0.05 \times 0.996 = 0.0498$$

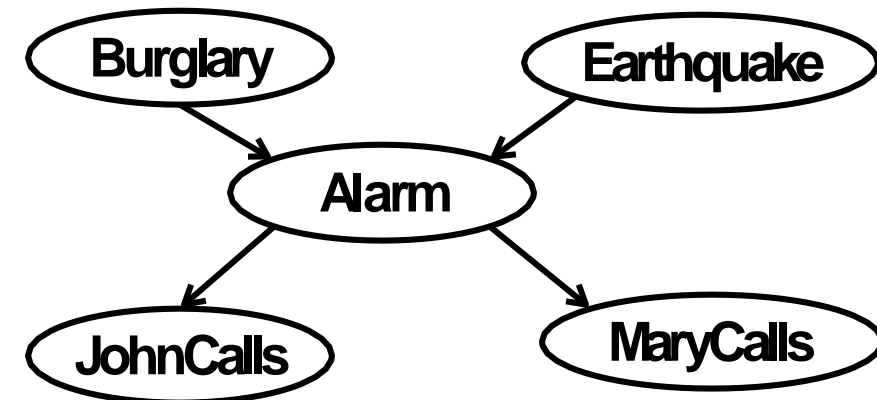
$$P(JE') = P(AJE') + P(A'JE') = 0.00175 + 0.0498 = 0.05155$$

B	E	P(A)
T	T	0.95
T	F	0.95
F	T	0.29
F	F	0.001

A	P(J)
T	0.90
F	0.05

A	P(M)
T	0.70
F	0.01

P(E)	P(B)
0.002	0.001



The joint probability distribution

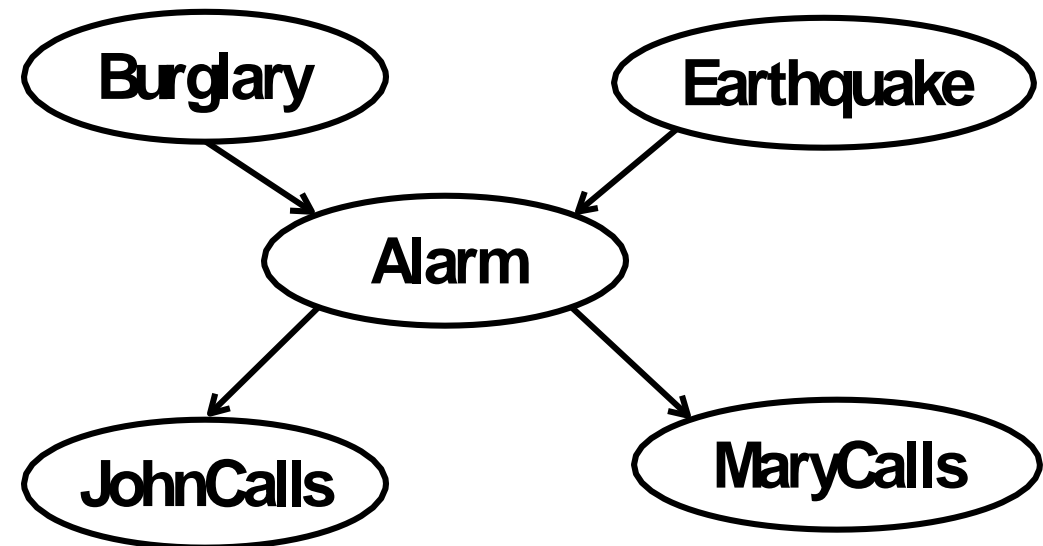
$$P(A | JE') = P(AJE') / P(JE') = 0.00175 / 0.05155 = 0.03$$

B	E	P(A)
T	T	0.95
T	F	0.95
F	T	0.29
F	F	0.001

A	P(J)
T	0.90
F	0.05

A	P(M)
T	0.70
F	0.01

P(E)	P(B)
0.002	0.001



The joint probability distribution

$$\begin{aligned}
 P(BJE') &= P(BJE'A) + P(BJE'A') \\
 &= P(J | ABE').P(ABE') + P(J | A'BE').P(A'BE') \\
 &= P(J | A).P(ABE') + P(J | A').P(A'BE') \\
 &= 0.9 \times 0.95 \times 0.001 \times 0.998 + 0.05 \times (1 - 0.95) \times 0.001 \times 0.998 \\
 &= 0.000856
 \end{aligned}$$

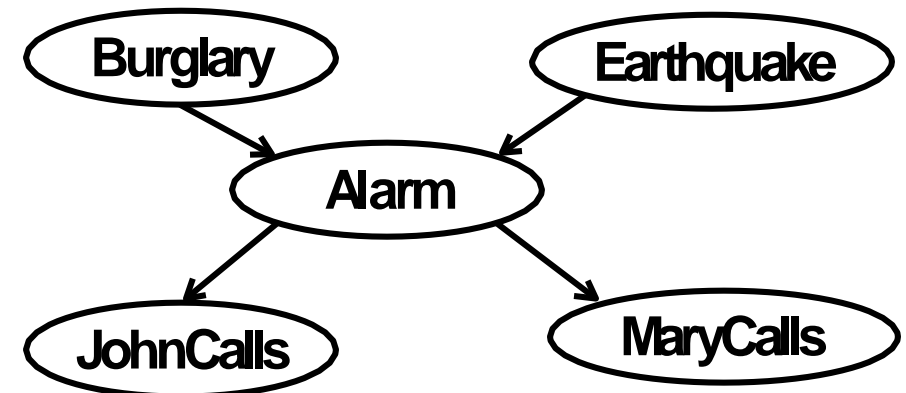
$$P(B | JE') = P(BJE') / P(JE') = 0.000856 / 0.05155 = 0.017$$

B	E	P(A)
T	T	0.95
T	F	0.95
F	T	0.29
F	F	0.001

A	P(J)
T	0.90
F	0.05

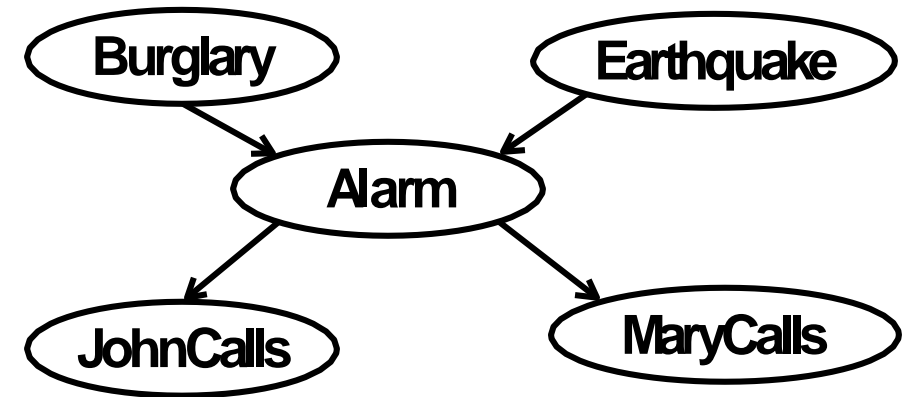
A	P(M)
T	0.70
F	0.01

P(E)	P(B)
0.002	0.001



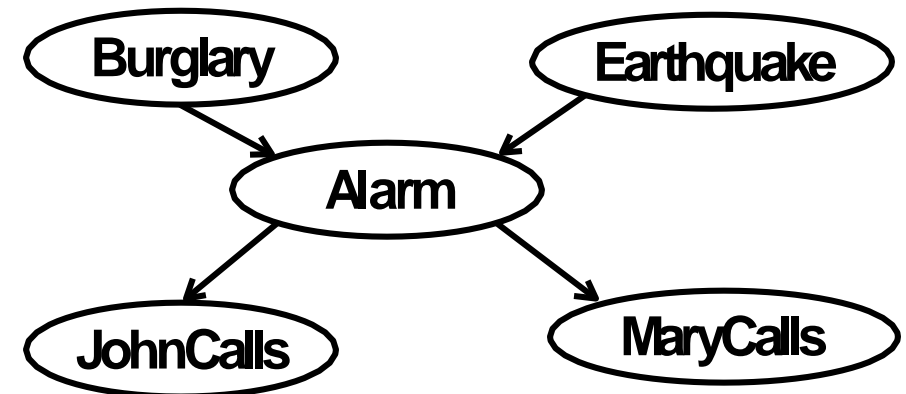
Inferences using belief networks

- Diagnostic inferences (from effects to causes)
 - Given that JohnCalls, infer that
$$P(\text{Burglary} \mid \text{JohnCalls}) = 0.016$$
- Causal inferences (from causes to effects)
 - Given Burglary, infer that
$$P(\text{JohnCalls} \mid \text{Burglary}) = 0.86$$
$$P(\text{MaryCalls} \mid \text{Burglary}) = 0.67$$



Inferences using belief networks

- Inter-causal inferences (between causes of a common effect)
 - Given Alarm, we have $P(\text{Burglary} \mid \text{Alarm}) = 0.376$
 - If we add evidence that Earthquake is true, then $P(\text{Burglary} \mid \text{Alarm} \wedge \text{Earthquake}) = 0.003$
- Mixed inferences
- Setting the effect JohnCalls to true and the cause Earthquake to false gives $P(\text{Alarm} \mid \text{JohnCalls} \wedge \neg \text{Earthquake}) = 0.003$



Exercise

Three candidates run for an election as a major in a city.

According to a public opinion poll, their chances to win are 0.25, 0.35 und 0.40.

The chances that they build a bridge after they have been elected are 0.60, 0.90 and 0.80.

What is the probability that the bridge will be built after the election?

Solution: Let $C, c \in \{1, 2, 3\}$, be the random variable indicating the winning candidate and $B, b \in \{t, f\}$, the random variable indicating whether the bridge will be built. Then the total probability that the bridge will be built is

$$P(B = t) = \sum_{c=1}^3 P(B = t|c)P(c) = 0.60 \times 0.25 + 0.90 \times 0.35 + 0.80 \times 0.40 = 0.785.$$

Exercise

On an airport all passengers are checked carefully.

Let T with $t \in \{0, 1\}$ be the random variable indicating whether somebody is a terrorist ($t = 1$) or not ($t = 0$).

Let A with $a \in \{0, 1\}$ be the variable indicating arrest.

A terrorist shall be arrested with probability $P(A = 1|T = 1) = 0.98$, a non-terrorist with probability $P(A = 1|T = 0) = 0.001$.

One in a lakh passengers is a terrorist, $P(T = 1) = 0.00001$.

What is the probability that an arrested person actually is a terrorist?