

AIFA: Reasoning Under Uncertainty

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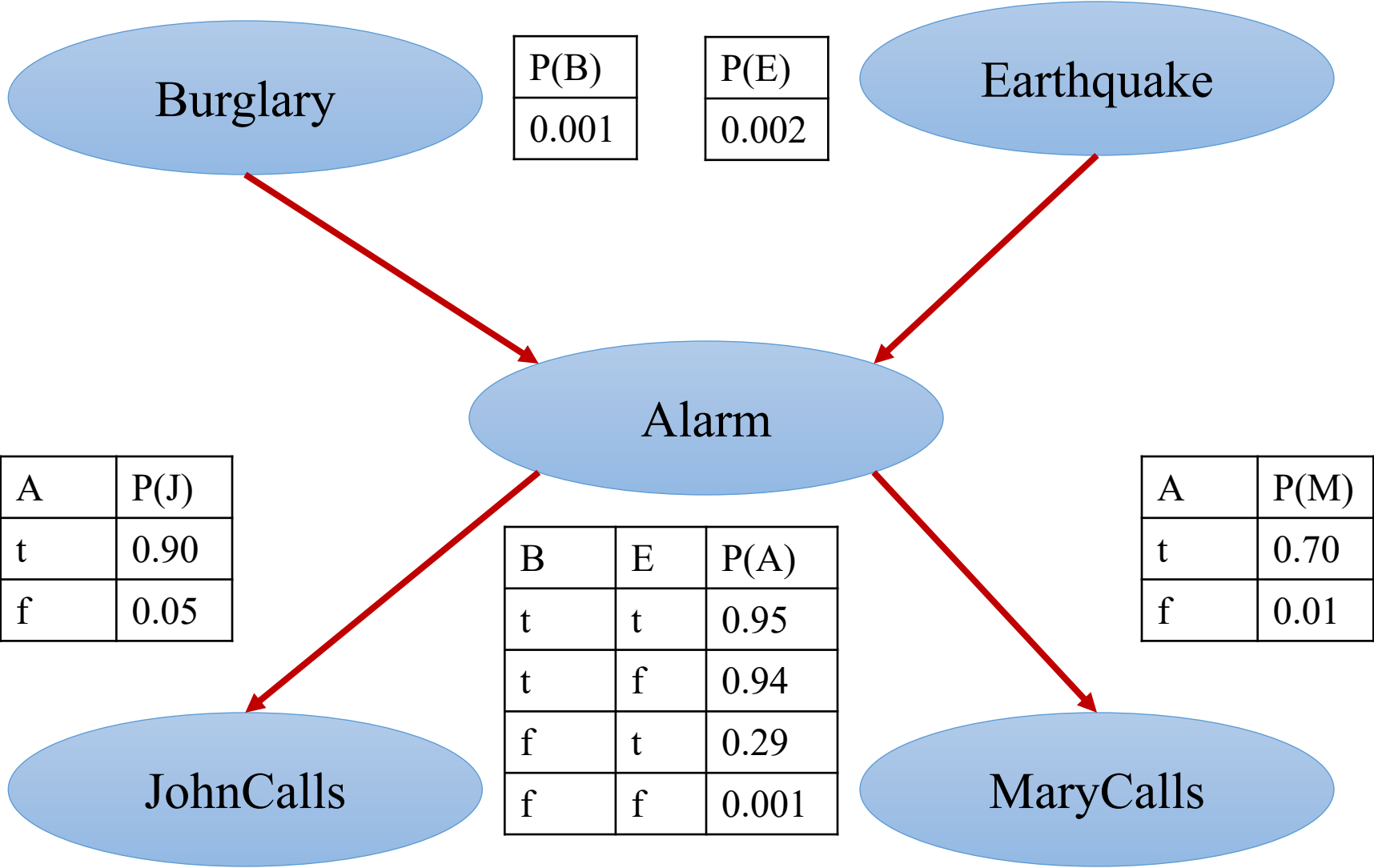
Bayesian Network

- A set of random variables makes up the nodes of the network
 - Variables may be discrete or continuous
- A set of directed links or arrows connects pairs of nodes
 - Arrows represent probabilistic dependence among variables
- An arrow from $X \rightarrow Y$ indicates X is parent of Y
- Each node X_i has a conditional probability distribution $P(X_i | \text{Parents}(X_i))$
 - Quantifies the effect of the parents on the node
- The graph has no directed cycles (DAG)

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
 - But sometimes misses the alarm altogether

Belief Network



Joint probability distribution

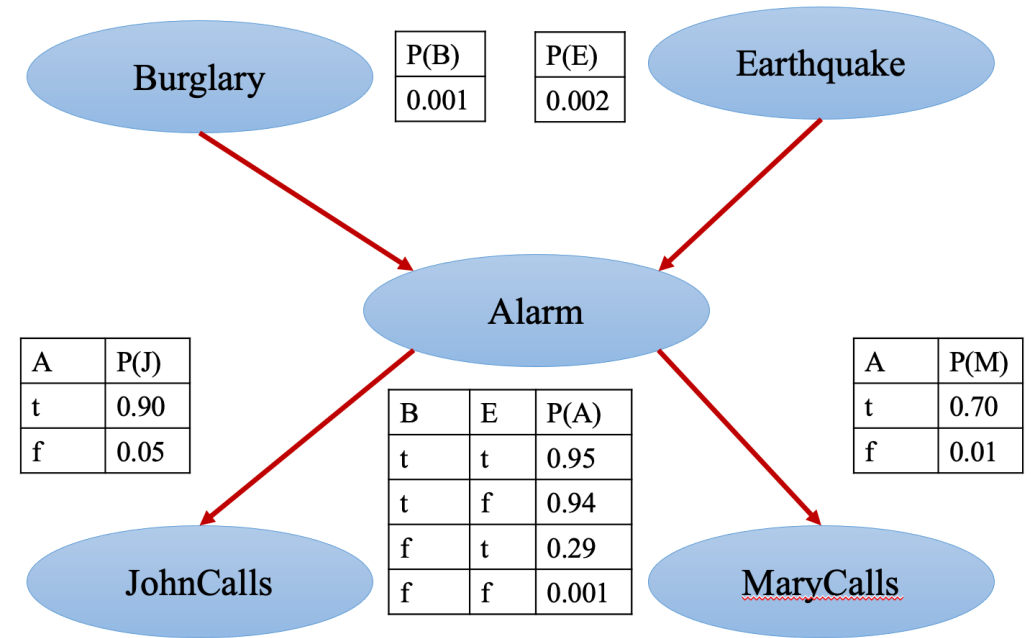
- $P(x_1, \dots, x_n) = \prod_{i=1}^n P(x_i | \text{Parents}(x_i))$

- $P(J \wedge M \wedge A \wedge \sim B \wedge \sim E)$

- $P(J|A) *$
- $P(M|A) *$
- $P(A|\sim B \wedge \sim E) *$
- $P(\sim B) *$
- $P(\sim E)$

- $P(J \wedge M \wedge A \wedge \sim B \wedge \sim E) = 0.9 \times 0.7 \times 0.001 \times 0.999 \times 0.998 = 0.00062$

- $P(J) = ?$



Conditional Independence

- $P(x_1, \dots, x_n) = P(x_n | x_{n-1}, \dots, x_1) P(x_{n-1}, \dots, x_1)$
 - $P(x_1, \dots, x_n) = P(x_n | x_{n-1}, \dots, x_1) P(x_{n-1} | x_{n-2}, \dots, x_1) \dots P(x_2 | x_1) P(x_1)$
 - $P(x_1, \dots, x_n) = \prod_{i=1}^n P(x_i | x_{i-1}, \dots, x_1)$
-
- The belief network represents conditional independence:
 - $P(x_i | x_i, \dots, x_1) = P(x_i | \text{Parents}(x_i))$

How to construct this network?

Conditional Independence

- $P(J, M, A, B, E) = P(J|M, A, B, E)P(M, A, B, E)$
- $P(J, M, A, B, E) = P(J|A)P(M|A, B, E)P(A, B, E)$
- $P(J, M, A, B, E) = P(J|A)P(M|A)P(A|B, E)P(B, E)$
- $P(J, M, A, B, E) = P(J|A)P(M|A)P(A|B, E)P(B)P(E)$

How does ordering matter?

Conditional Independence

Conditional Independence Relation

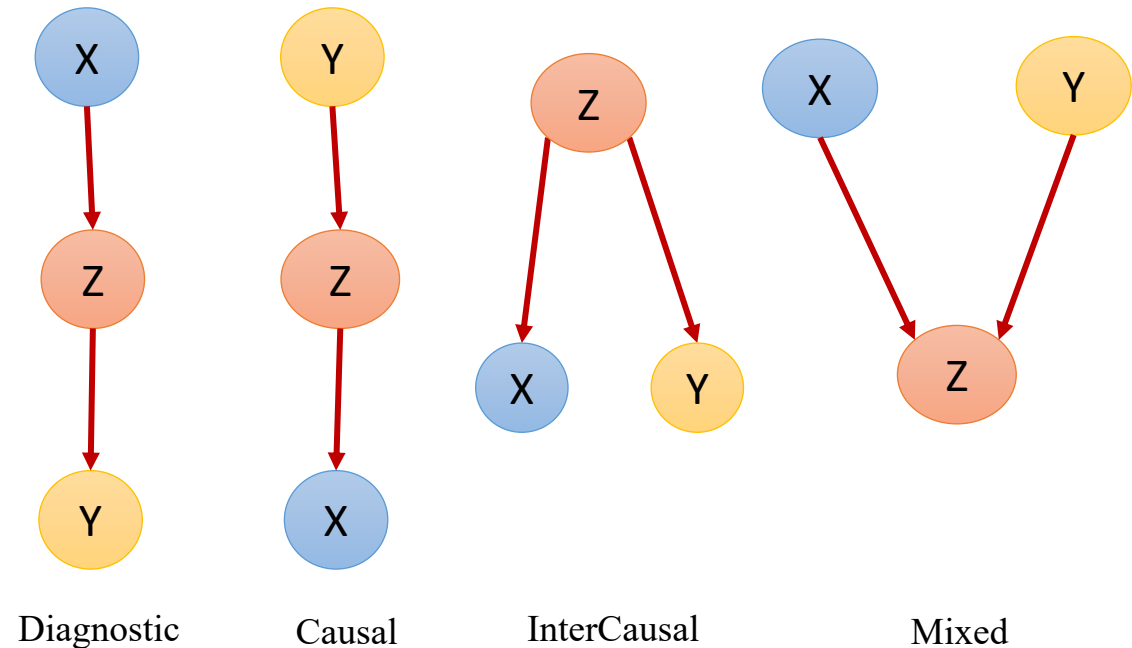
- If every undirected path from a node in X to a node in Y is d-separated by a given set of evidence nodes E
 - X and Y are conditionally independent given E
- A set of nodes E d-separates two set of nodes X and Y if every undirected path from a node in X to a node in Y is blocked given E

Conditional Independence Relation

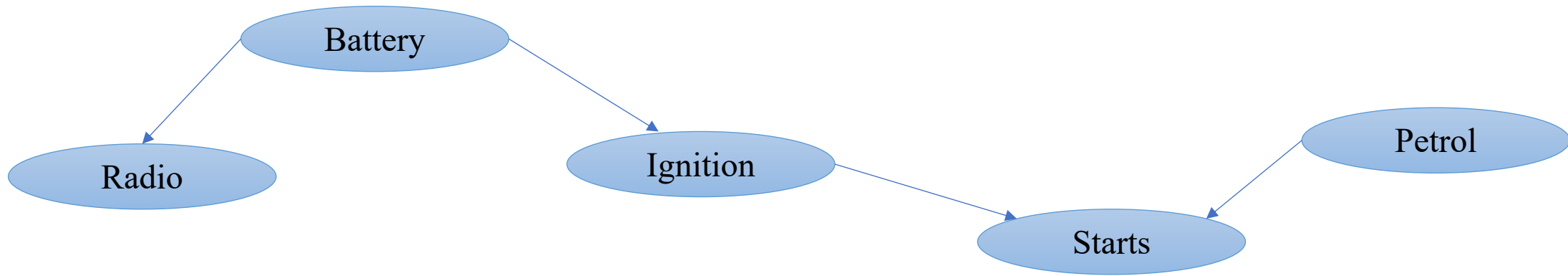
- A path is blocked given a set of nodes E if there is a node Z on the path for which one of three conditions hold:
 - Z is in E and Z has one arrow on the path leading in and one arrow out
 - Z is in E and Z has both path arrows leading out
 - Neither Z nor any descendent of Z is in E and both path arrows lead into Z

Conditional Independence Relation: Summary

- A path is blocked given a set of nodes E if there is a node Z on the path for which one of three conditions hold:
 - Z is in E and Z has one arrow on the path leading in and one arrow out
 - Z is in E and Z has both path arrows leading out
 - Neither Z nor any descendent of Z is in E and both path arrows lead into Z
- If every undirected path from a node in X to a node in Y is **d-separated** by a given set of evidence nodes E
 - X and Y are conditionally independent given E
- A set of nodes E **d-separates** two set of nodes X and Y if every undirected path from a node in X to a node in Y is **blocked** given E

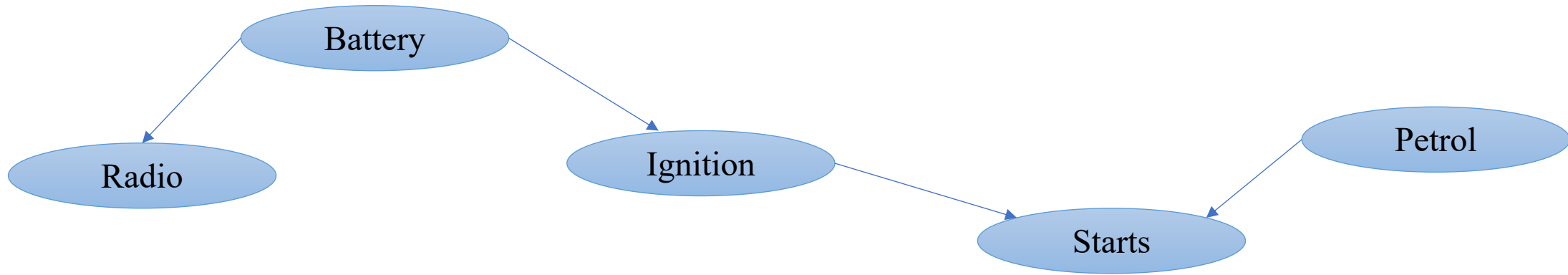


Conditional Independence Relation



- Whether there is petrol and whether the radio plays are independent given evidence about whether the ignition takes place
- Petrol and Radio are independent if it is known whether the battery works

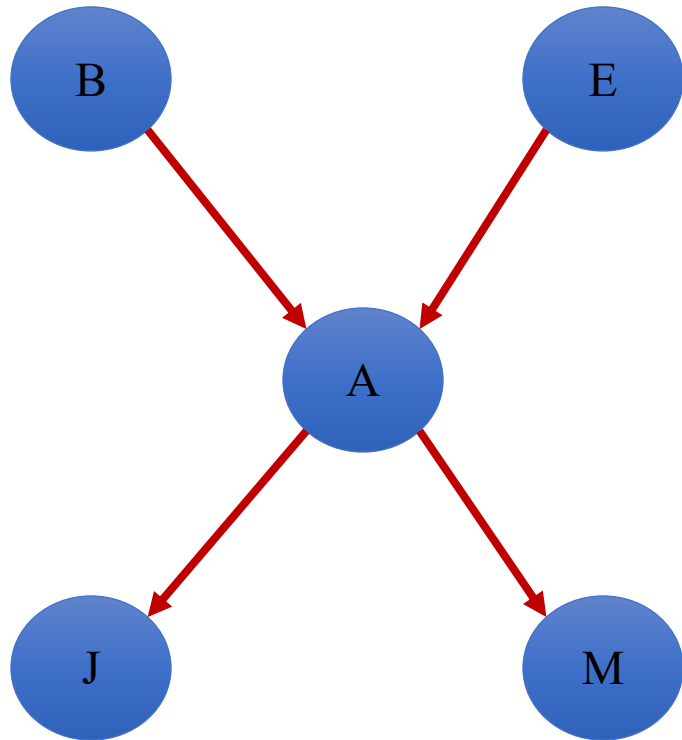
Conditional Independence Relation



- Petrol and Radio are independent given no evidence at all
- But they are dependent given evidence about whether the car starts
- If the car does not start, then radio works is an increase evidence that car is out of petrol

Bayesian Network: Topological Semantics

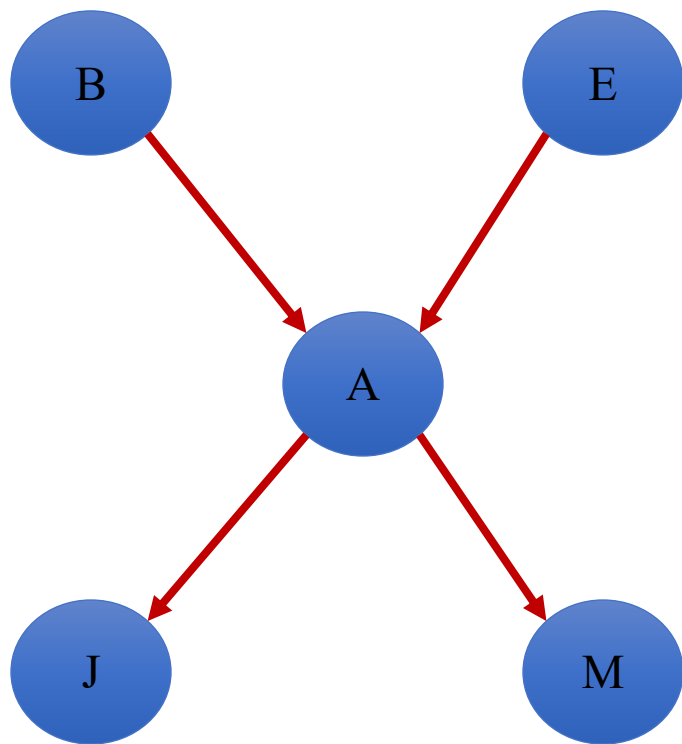
- A node is conditionally independent of its non-descendants, given its parents



JohnCalls is independent of Burglary and Earthquake given the value of Alarm

Bayesian Network: Topological Semantics

- A node is conditionally independent of all other nodes in the network, given its parents, children, and children's parents
 - Markov Blanket



Burglary is independent of JohnCalls and MaryCalls given the value of Alarm and Earthquake

Thank You