

# Independence and Bayesian Networks

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Lecture 11

Readings: RN 13.4 & 14.2. PM 8.2 & 8.3.

# Outline

Learning Goals

Unconditional and Conditional Independence

Examples of Bayesian Networks

Why Bayesian Networks

Representing the Joint Distribution

Independence in Three Key Structures

Revisiting the Learning goals

# Learning Goals

By the end of the lecture, you should be able to

- ▶ Given a probabilistic model, determine if two variables are unconditionally independent, or conditionally independent given a third variable.
- ▶ Give examples of deriving a compact representation of a joint distribution by using independence assumptions.
- ▶ Describe components of a Bayesian network.
- ▶ Compute a joint probability given a Bayesian network.
- ▶ Explain the independence relationships in the three key structures.

Learning Goals

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# (Unconditional) Independence

Definition ((unconditional) independence)

$X$  and  $Y$  are (unconditionally) independent iff

$$P(X|Y) = P(X)$$

$$P(Y|X) = P(Y)$$

$$P(X \wedge Y) = P(X)P(Y)$$

Learning  $Y$  does NOT influence your belief about  $X$ .

# Conditional Independence

## Definition (conditional independence)

$X$  and  $Y$  are conditionally independent given  $Z$  if

$$P(X|Y \wedge Z) = P(X|Z).$$

$$P(Y|X \wedge Z) = P(Y|Z).$$

$$P(Y \wedge X|Z) = P(Y|Z)P(X|Z).$$

Learning  $Y$  does NOT influence your belief about  $X$  if you already know  $Z$ .

## CQ: Deriving a compact representation

**CQ:** Consider a model with three random variables,  $A, B, C$ .

1. What is the minimum number of probabilities required to specify the joint distribution?
  
  
  
  
  
  
  
  
  
  
2. Assume that  $A, B$ , and  $C$  are independent.  
What is the minimum number of probabilities required to specify the joint distribution?

## CQ: Deriving a compact representation

**CQ:** Consider a model with three random variables,  $A, B, C$ .

1. What is the minimum number of probabilities required to specify the joint distribution?
  
  
  
  
  
  
  
  
  
  
2. Assume that  $A$  and  $B$  are conditionally independent given  $C$ . What is the minimum number of probabilities required to specify the joint distribution?



Learning Goals

Unconditional and Conditional Independence

Examples of Bayesian Networks

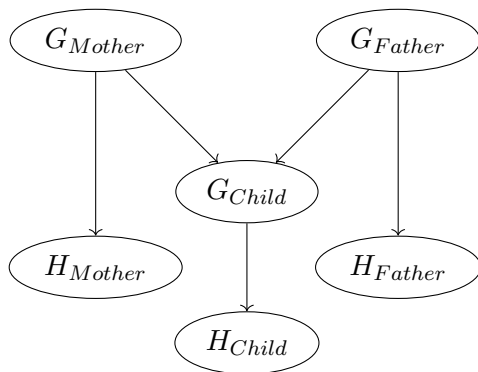
Why Bayesian Networks

Representing the Joint Distribution

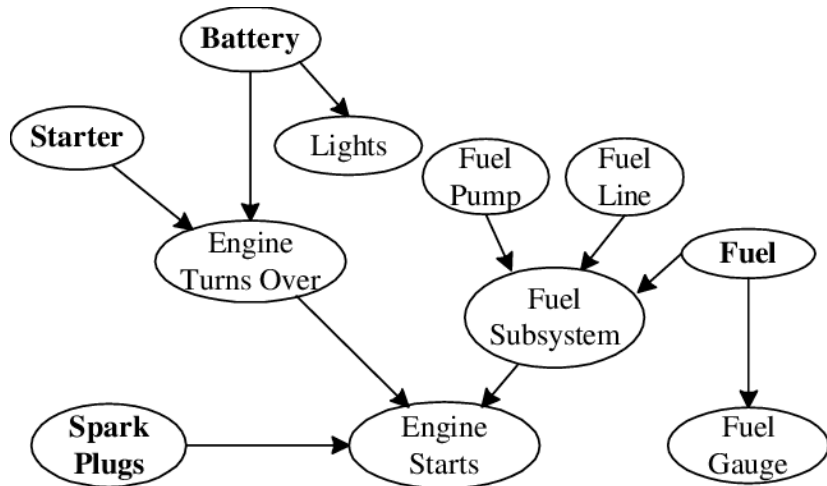
Independence in Three Key Structures

Revisiting the Learning goals

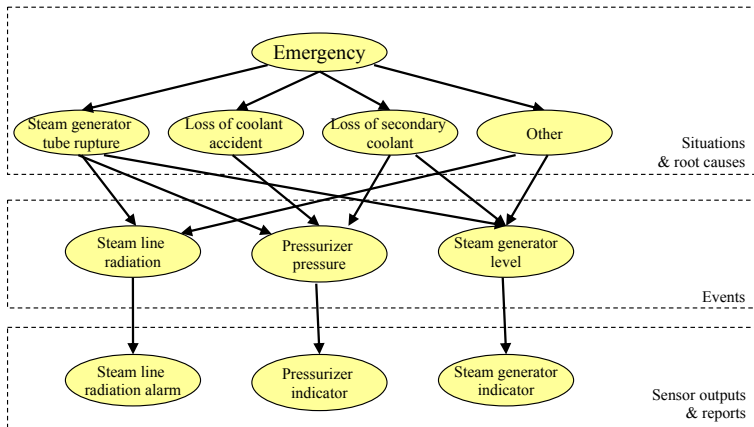
# Inheritance of Handedness



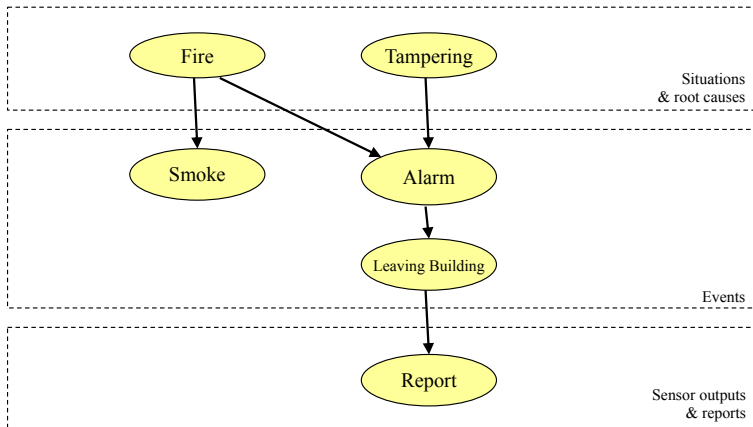
# Car Diagnostic Network



## Example: Nuclear power plant operations

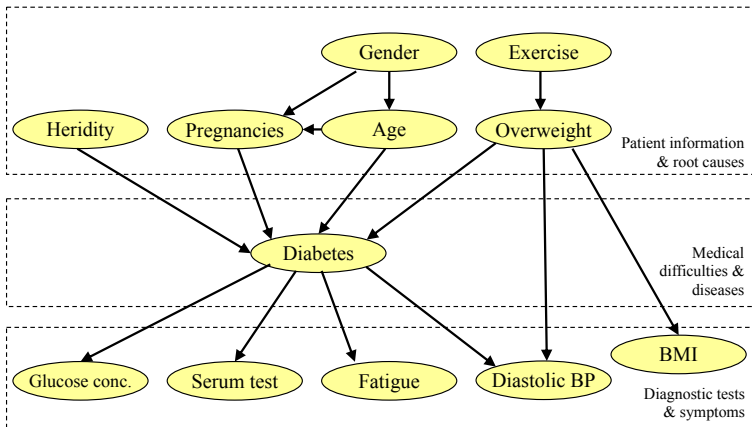


## Example: Fire alarms



Report: “report of people leaving building because a fire alarm went off”

## Example: Medical diagnosis of diabetes



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# Why Bayesian Networks?

A probabilistic model of the Holmes scenario:

- ▶ The random variables:  
Earthquake, Radio, Burglary, Alarm, Watson, and Gibbon.
- ▶ # of probabilities in the joint distribution:  $2^6 = 64$ .
- ▶ For example,  
 $P(E \wedge R \wedge B \wedge A \wedge W \wedge G) = ?$   
 $P(E \wedge R \wedge B \wedge A \wedge W \wedge \neg G) = ?$   
... etc ...

We can compute any probability using the joint distribution, but

- ▶ Quickly become intractable as the number of variables grows.
- ▶ Unnatural and tedious to specify all the probabilities.

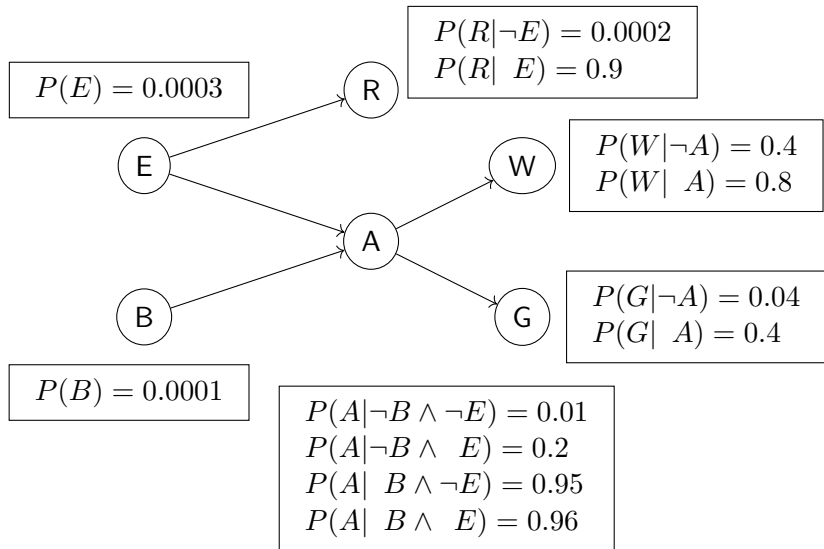


# Why Bayesian Networks?

## A Bayesian Network

- ▶ is a **compact** version of the joint distribution, and
- ▶ takes advantage of the **unconditional and conditional independence** among the variables.

# A Bayesian Network for the Holmes Scenario



# Bayesian Network

A Bayesian Network is a directed acyclic graph.

- ▶ Each node corresponds to a random variable.
- ▶  $X$  is a parent of  $Y$  if there is an arrow from node  $X$  to node  $Y$ .
- ▶ Each node  $X_i$  has a conditional probability distribution  $P(X_i | Parents(X_i))$  that quantifies the effect of the parents on the node.

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# The Semantics of Bayesian Networks

Two ways to understand Bayesian Networks:

- ▶ A representation of the joint probability distribution
- ▶ An encoding of the conditional independence assumptions

## Representing the joint distribution

We can compute each joint probability using the following formula.

$$P(X_n \wedge \cdots \wedge X_1) = \prod_{i=1}^n P(X_i | Parents(X_i))$$

# Representing the joint distribution

**Example:** What is the probability that

- ▶ The alarm has sounded,
- ▶ Neither a burglary nor an earthquake has occurred,
- ▶ Both Watson and Gibbon call and say they hear the alarm,  
and
- ▶ There is no radio report of an earthquake?

## CQ: Calculating the joint probability

**CQ:** What is the probability that

- ▶ NEITHER a burglary NOR an earthquake has occurred,
- ▶ The alarm has NOT sounded,
- ▶ NEITHER of Watson and Gibbon is calling, and
- ▶ There is NO radio report of an earthquake?

(A) 0.5699

(B) 0.6699

(C) 0.7699

(D) 0.8699

(E) 0.9699



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# Burglary, Alarm and Watson



## CQ Unconditional Independence

**CQ:** Are Burglary and Watson independent?



- (A) Yes
- (B) No

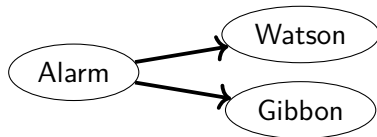
## CQ: Conditional Independence

**CQ:** Are Burglary and Watson conditionally independent given Alarm?



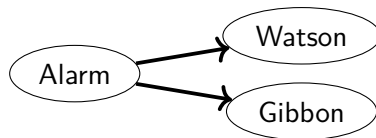
- (A) Yes
- (B) No

# Alarm, Watson and Gibbon



## CQ Unconditional Independence

**CQ:** Are Watson and Gibbon independent?

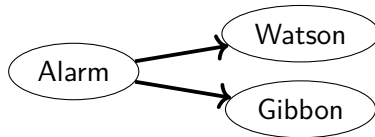


(A) Yes

(B) No

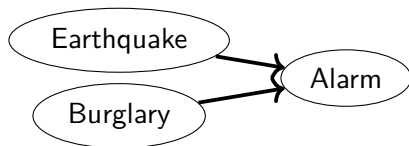
## CQ Conditional Independence

**CQ:** Are Watson and Gibbon conditionally independent given Alarm?



- (A) Yes
- (B) No

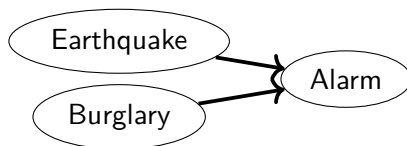
# Earthquake, Burglary, and Alarm





## CQ Unconditional Independence

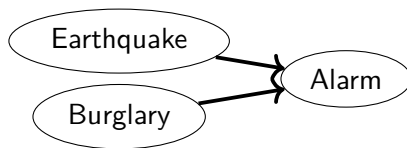
**CQ:** Are Earthquake and Burglary independent?



- (A) Yes
- (B) No

## CQ: Conditional Independence

**CQ:** Are Earthquake and Burglary conditionally independent given Alarm?



- (A) Yes
- (B) No

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