

## Lecture 10: October 28

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## 10.1 Recap: Probability

### 10.1.1 Axioms

$$1. 0 \leq P(a) \leq 1$$

$$2. P(a \cup b) = P(a) + P(b) - P(a \cap b)$$

$$3. P(True) = 1, P(False) = 0$$

### 10.1.2 Bayes' theorem

$$P(a|b) = \frac{P(b|a)P(a)}{P(b)}$$

### 10.1.3 Independence

If  $P(a|b) = P(a)$ , then  $a$  is independent of  $b$ .

A special case of independence is if  $P(a|b, c) = P(a|b)$ , then  $a$ , given  $b$ , is **conditionally independent of**  $c$ .

## 10.2 Bayesian Networks

### 10.2.1 What

When drawing a probability table for a event  $c$ , if it is dependent on event  $a$  and  $b$ , then the table will look like the following example:

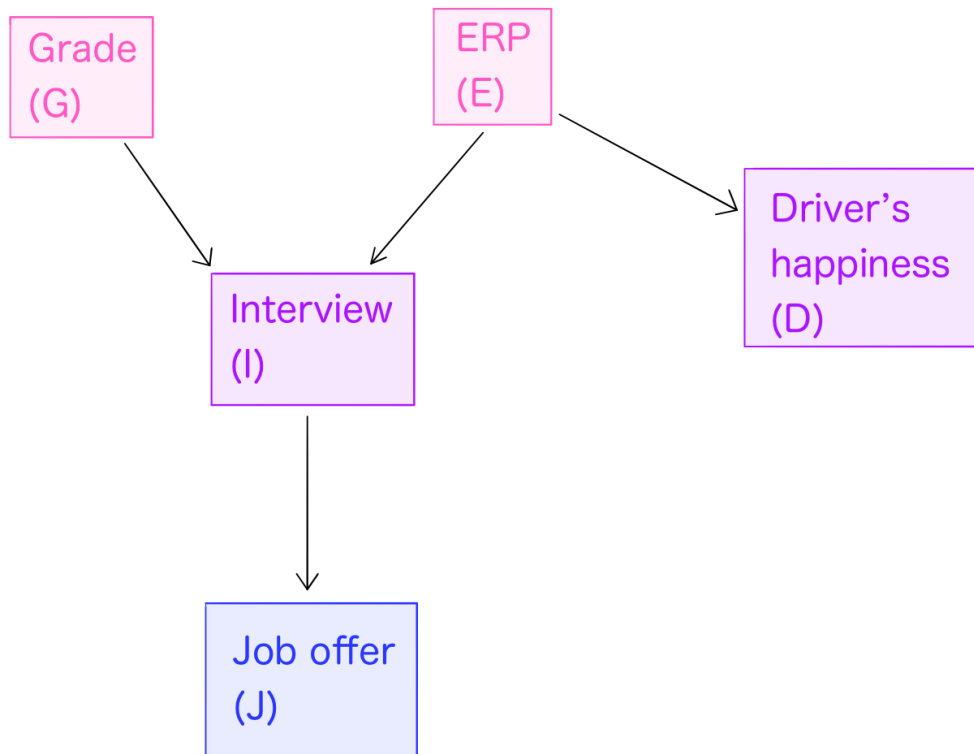
a	b	c
T	T	$P(c a,b)$
T	F	$P(c a,!b)$
F	T	$P(c !a,b)$
F	F	$P(c !a,!b)$

Since  $c$  is dependent on 2 events, therefore its probability table has 4 rows. For an event that is dependent on  $k$  other events, its table will have  $2^k$  rows.

In real world, there are way too many events. When drawing a probability table for a single event, it is impossible to involve all other events because it will result in a table of  $2^n$  where  $n$  is a huge number.

Bayesian Networks abstract causal relationships from the real world and form a **directed acyclic graph**, where each directed edge represents a '*ancestor-descendent*' or '*depended-depends*' relationship.

The example mentioned in lecture is attached below:



### 10.2.2 Axiom

Each node in a Bayesian Network is independent of its non-descendants, given its parents.

### 10.2.3 How

Based on the axiom,  $P(I|G, E, D) = P(I|G, E)$  because  $D$  is a non-descendant of  $I$ .

Suppose we want to know  $P(G, E, I, D, J)$ , we can use the axiom smartly to reduce the complexity:

$$\begin{aligned}
 P(G, E, I, D, J) &= P(J, I, D, G, E) \\
 &= P(J|I, D, G, E)P(I|D, G, E)P(D|G, E)P(G|E)P(E) \\
 &= P(J|I)P(I|G, E)P(D|E)P(G)P(E)
 \end{aligned} \tag{10.1}$$

Now, we only need probability tables of 10 rows.

Note that to best exploit the axiom, we need to arrange the events in **topological order**, this can be easily done by recursively picking the leaf node.