

# 1 Basic probability

## 1.0.1 Inference by enumeration

$$P(\text{Effect} \mid \text{Cause}) = \frac{P(\text{Effect} \wedge \text{Cause})}{P(\text{Cause})} = \alpha P(\text{Effect}, \text{Cause})$$

$\alpha$  = Normalization constant

## 1.0.2 Bayes theorem

$$P(\text{Cause} \mid \text{Effect}) = \frac{P(\text{Effect} \mid \text{Cause})P(\text{Cause})}{P(\text{Effect})} = \alpha P(\text{Effect} \mid \text{Cause})P(\text{Cause})$$

# 2 Bayesian Networks

## 2.1 Flow of probabilistic influence

- **X direct cause** of Y:  $X \rightarrow Y$
- **X direct effect** of Y:  $X \leftarrow Y$
- **Causal trail** from X to Y:  $X \rightarrow Z \rightarrow Y$
- **Evidential trail** from X to Y:  $X \leftarrow Z \leftarrow Y$
- **Common cause**:  $X \leftarrow Z \rightarrow Y$
- **Common effect**:  $X \rightarrow Z \leftarrow Y$

### 2.1.1 Active trail

Let  $Z$  be a subset of observed variables. The trail  $X_{i-1} \rightleftharpoons X_i \rightleftharpoons X_{i+1}$  is active given  $Z$  if

- $\forall X_{i-1} \rightarrow X_i \leftarrow X_{i+1}$ ,  $X_i$  or one of its descendants are in  $Z$
- No other node along the trail is in  $Z$

### 2.1.2 Direct separation

To determine if  $P \models (X \perp Y \mid Z)$  ( $X$  and  $Y$  are independent given  $Z$ ):

1. Traverse the graph bottom-up marking all nodes in  $Z$  or having descendants in given  $Z$
2. Traverse the graph from  $X$  to  $Y$ , stopping if we get to a blocked node\*
3. If we can't reach  $Y$ , then  $X$  and  $Y$  are independent

\*A node is blocked if either the middle of an unmarked v-structure, or in  $Z$ , but not both

## 2.2 Reasoning Patterns

- **Causal** (or predictive) reasoning:  $P(\text{Descendant} \mid \text{Ancestor})$
- **Evidential** (or explanatory) reasoning:  $P(\text{Ancestor} \mid \text{Descendant})$
- **Intercausal** reasoning (or explaining away):  $P(Y \mid X)$  where  $X$  is connected to  $Y$  via an active trail which passes through a common cause and/or a common effect