

Chapter 1 Bayesian Network

Theorem 1.1

Let $\{X_i\}_{i=1}^n$ are random variables, then

$$P(\cap_{i=1}^n X_i) = \prod_{j=1}^n P(X_j | \cap_{i=1}^{j-1} X_i)$$

consider the following probability directed acyclic graph:

$$\begin{aligned} P(X_1, X_2, X_3, X_4) &= P(X_1 | X_2, X_3, X_4) \times P(X_2 | X_3, X_4) \times P(X_3 | X_4) \times P(X_4) \\ &= P(X_1 | X_3, X_4) \times P(X_2 | X_3) \times P(X_3) \times P(X_4) \end{aligned}$$

Theorem 1.2

Let $G = \langle V, E \rangle$ be a probability directed acyclic graph. where $V = \{X_i\}_{i=1}^n$.

$$P(\cap_{i=1}^n X_i) = \prod_{i=1}^n P(X_i | \text{parent}(X_i))$$

Proof. let's try a constructive proof, at first, we have known that $P(\cap_{i \in [1, n]} X_i) = P(\cap_{i \in \text{permute}([1, n])} X_i)$

$$P(\cap_{i=1}^n X_i) = P(X_1 | X_2, X_3, \dots, X_n) \times P(X_2, X_3, \dots, X_n)$$

therefore $\text{Parent}(X_1) \subseteq \{X_i\}_{i=2}^n$ hold.

then consider $P(X_2, X_3, \dots, X_n) = P(X_2 | X_3, \dots, X_n) \times P(X_3, \dots, X_n)$, Since G is a directed acyclic graph, there are only three cases to decompose it into the one we want:

Case1 $X_1 \notin \text{Parent}(X_2)$ and $X_2 \notin \text{Parent}(X_1)$

In this case, it's trivial

Case2 $X_1 \notin \text{Parent}(X_2)$ and $X_2 \in \text{Parent}(X_1)$

All right, we can continue to decomposition $P(X_2, \dots, X_n)$ to $P(X_2 | X_3, \dots, X_n)$, because G is a DAG (directed acyclic graph).

Case3 $X_1 \in \text{Parent}(X_2)$ and $X_2 \notin \text{Parent}(X_1)$

Back to previous step and replace $P(X_2, X_1, X_3, \dots, X_n)$ with $P(X_1, X_2, X_3, \dots, X_n)$

Continue with the above process until the end of the decomposition and we have got

$$P(\cap_{i=1}^n X_i) = \prod_{i=1}^n P(X_i | \text{parent}(X_i))$$

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