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Note Title

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Then (The chain rule for Bayesian networks)

Let  $G = (V, E, P)$  be a BN. Then,

$P(V) = \prod_{v \in V} P(v | c(v))$ . [The joint prob. of all vars in a BN  
is the product of its CPTs.]

~~conditional probability table~~

Let  $|V| = n$ . Reorder (if necessary) the variables

in  $V$ , so that  $\langle v_1, v_2, \dots, v_{n-1}, v_n \rangle$  is a

reverse topological order for the DAG  $(V, E)$ .

$$\begin{aligned}
 P(v) &= P(v_1, v_2, \dots, v_{n-1}, v_n) = (\text{by the fundamental rule}) \\
 P(v) &= P(v_1 | v_2, \dots, v_{n-1}, v_n) P(v_2, \dots, v_{n-1}, v_n) = \\
 &= (\text{by the definition of Bayesian network}) = \\
 &= P(v_1 | c(v_1)) P(v_2, \dots, v_{n-1}, v_n) = (\text{by the f.r.u.l}) = \\
 &= P(v_1 | c(v_1)) P(v_2 | v_3, \dots, v_n) P(v_3, \dots, v_n) = (\text{by the defn. of BN}) = \\
 &= P(v_1 | c(v_1)) P(v_2 | c(v_2)) P(v_3, \dots, v_n) =
 \end{aligned}$$

$$\cdots = P(r_1 | c(r_1)) P(r_2 | c(r_2)) \cdots P(r_{n-1} | c(r_{n-1})) P(r_n)$$