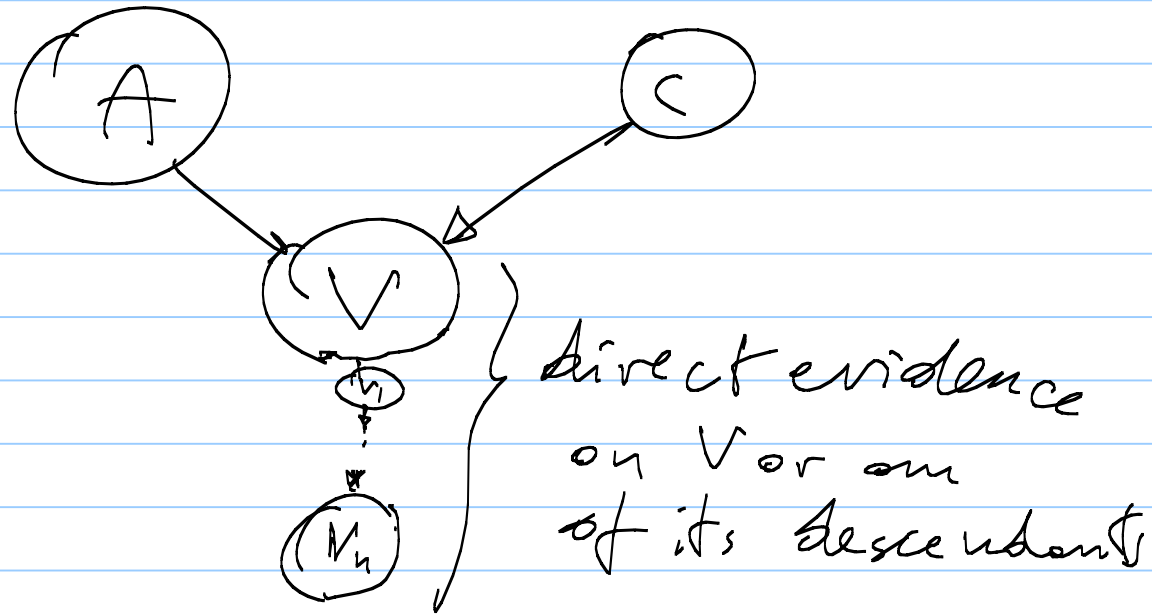


Turn in HW1 on Friday,
email before class or bring in a hard copy



Evidence is a set of findings.
Each finding is a statement that a variable is in

a particular state.

$e_i = \{ \text{Start} = u \}$ is evidence consisting of a single finding

First compute $P(\text{Religion}, \# \text{Children}, e)$ by setting to 0 the entries in $P(\text{Religion}, \# \text{Children})$ for which $\# \text{Children} \neq 3$, then normalize by the sum of all entries in $P(\text{Religion}, \# \text{Children}, e)$.

This works b/c the normalization constant is

$$\sum_{\# \text{Children} = 3} P(\# \text{Children} = 3)$$

Chain rule for Bayesian networks

This rule (a theorem) is an identity relating the joint prob. of all vars in a BN with the CPT (conditional probability tables).

1. Order the vars in reverse topological order

Topological order of a DAG (directed acyclic graph)

$G = (V, E)$ is an ordering $\langle v_1, \dots, v_n \rangle$ s.t.

$\{v_1, \dots, v_n\} = V$, $i < j, 1 \leq i, j \leq n$ & if i is a predecessor of j then $i < j$.

if j is a descendant of i

Please review this defn. (topological order).

There is a linear-time algorithm (due to Knuth) to compute a topological order in a DAG.

2. Apply the fundamental rule with the last var.
conditioned by all others

3. Apply d-separation

4. Repeat 2+3