

Bayesian Belief Nets

Conditional independence assumptions only apply to a subset of variables.

$$\text{ex: } P(X=x_1 | Y=y_1, Z=z_1) = P(X=x_1 | Z=z_1)$$

↳ we can get rid of this

- Assumption: X is conditionally independent of Y given Z if distribution governing X is independent of value of Y given a value for Z , if above equation holds.

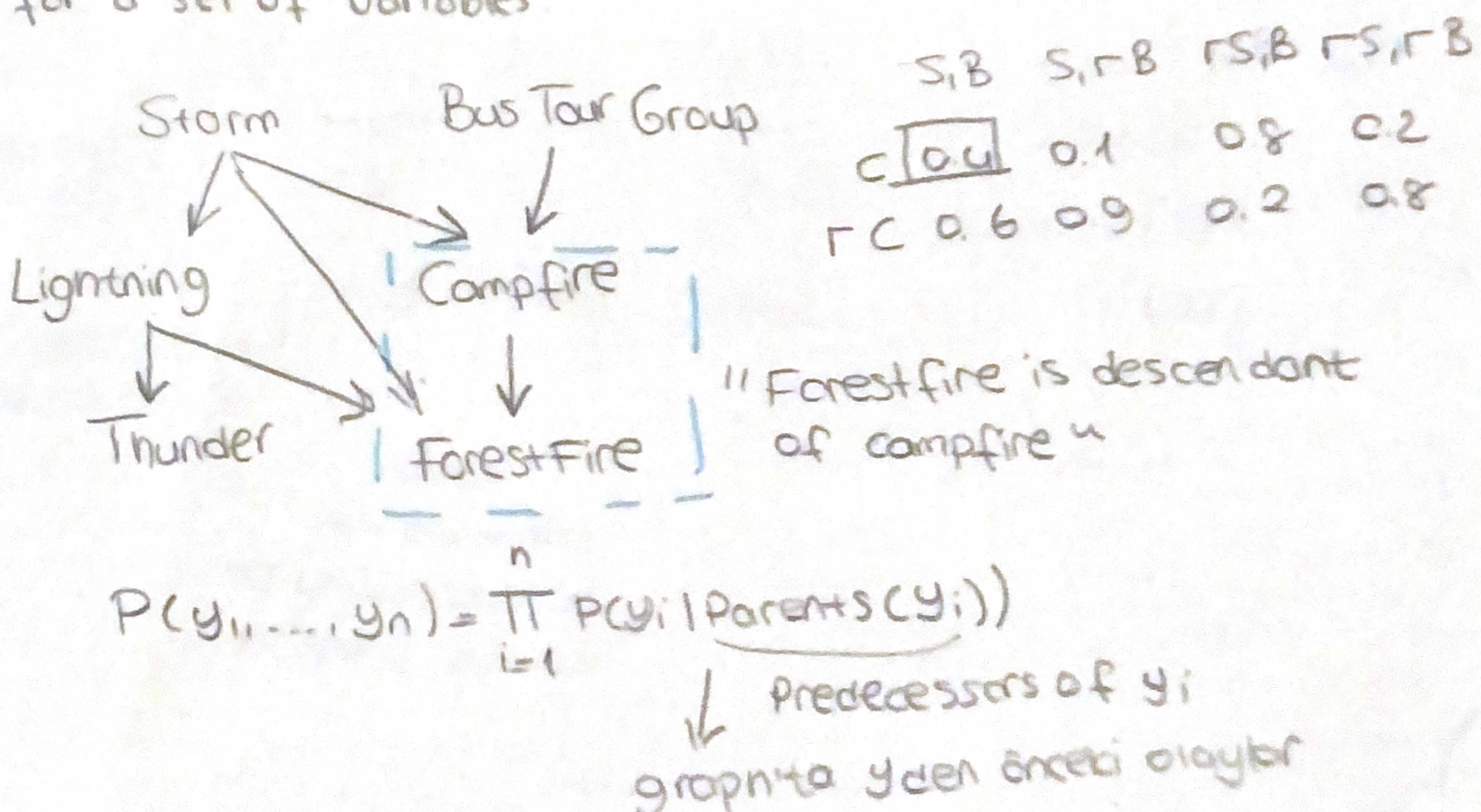
$$P(X|Y, Z) = P(X|Z)$$

↳ X ve Y independent so Y 'den kurtulmazsin

- Naive Bayesian independence assumption: saysinde:

$$\begin{aligned} P(A_1, A_2 | V) &= P(A_1 | A_2, V) P(A_2 | V) \\ &= P(A_1 | V) P(A_2 | V) \end{aligned}$$

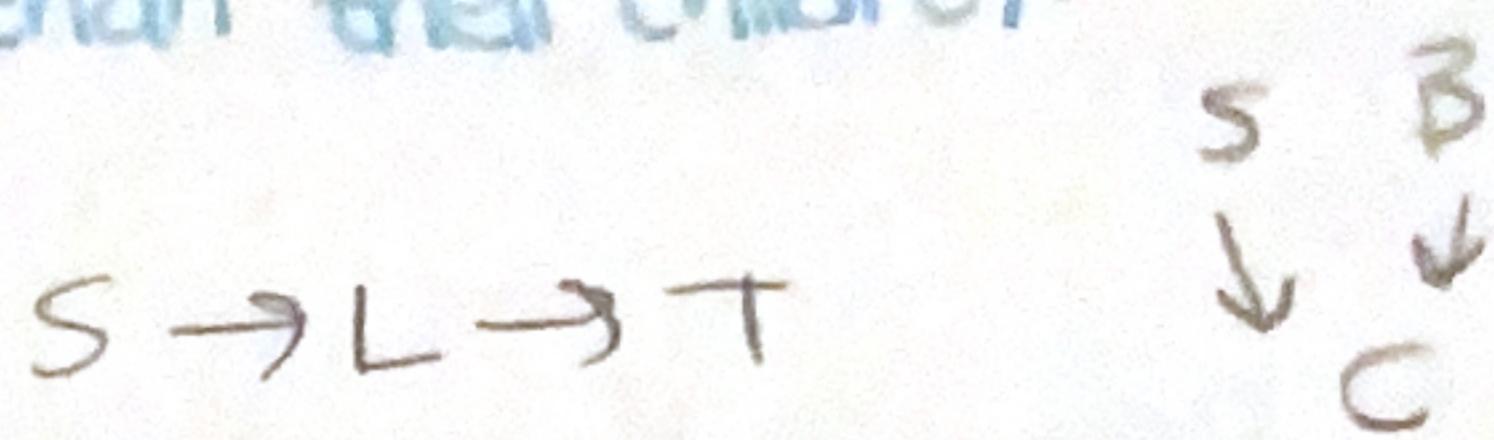
A Belief Network represents joint probability distribution for a set of variables



Campfire is independent of it's nondescendants Lightning & Thunder.

$$P(\text{Campfire} = \text{True} | \text{Storm} = \text{True}, \text{Bus tour group} = \text{True}) = \underline{\underline{0.6}}$$

In a Bayesian Net, we can number each node E_i . parents will have smaller indexes than their children.



- $P(\text{Campfire} | S, \neg B) = 0.1$
- $P(\text{Thunder, Lightning, Storm}) = P(T|L) P(L|S) P(S)$
- $P(\text{Campfire}) = P(C|S, B) P(S) P(B) + P(C|S, \neg B) P(S) P(\neg B)$
 $+ P(C|\neg S, B) P(\neg S) P(B) + P(C|\neg S, \neg B) P(\neg S) P(\neg B)$

- $P(B) = \sum_{i=1}^n P(B|A_i)$ where A_i add up to 1

- $P(C|S) = ?$ (we don't know about bus tour group?)

$$P(C|S) = P(C|S, B) P(B) + P(C|S, \neg B) P(\neg B) \rightarrow \text{Storm=True}$$

iterate over

- If most of the variables are unknown the complexity becomes exponential (2^n)
- We can use Monte Carlo simulation for this.

ex: $P(S) = 0.9$

$P(B) = 0.5$

$P(C|B, S) = 0.4$

$P(C) = ?$

<u>$P(S)$</u>	<u>$P(B)$</u>	<u>$P(C S, B)$</u>
$S \rightarrow T$	$B \rightarrow T$	$C \rightarrow F$
$S \rightarrow T$	$B \rightarrow F$	$B \rightarrow F$
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$S \rightarrow F$	$B \rightarrow T$	$C \rightarrow T$

} Generate cases

consider these cases
ratio obtained will be approximation to the campfire