

1.1

$$P(A, B, C, D, E, F) = P(A)P(B|A, F)P(C|B)P(D|C)P(E)P(F)$$

1.2

a. $A \perp\!\!\!\perp B \leftrightarrow A, B$ independent **F**

Because $P(B|A, F) \neq P(B) \leftrightarrow$ Value of B depends on value of A

b. $A \perp\!\!\!\perp F \leftrightarrow A, F$ independent **T**

Because $P(A, F) = P(A)P(F) \leftrightarrow$ nodes A, F (\leftrightarrow corresponding r.v.) are not related to each other

c. $A \perp\!\!\!\perp C | \{B, E\} \leftrightarrow A, C$ independent given B and E **T**

$$= \underbrace{A \perp\!\!\!\perp C | B} \cup \underbrace{A \perp\!\!\!\perp C | E}$$

T because **T** Since E is hidden on anything once B fixed

$$\Rightarrow A \perp\!\!\!\perp C$$

Formally

$$A \perp\!\!\!\perp C | \{B, E\} \leftrightarrow P(A, C | B, E) \stackrel{\text{Bayes}}{=} \frac{P(A, C, B, E)}{P(B, E)} = \frac{P(A)P(C|B)P(B|A)P(E)}{P(B|A)P(E)}$$

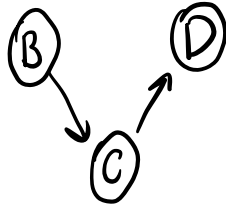
$$= P(A)P(C|B) \leftrightarrow \text{Factorizes}$$

In poor words: Fixed the var in the middle, no one gives a s**t about the var on the other side of the fixed one

$$d. F \perp\!\!\!\perp D | B \leftrightarrow P(F, D | B) = \frac{P(F, D, B)}{P(B)} = P(F) \frac{P(D|B)}{P(B)}$$

$$= P(F)P(D|B) \Rightarrow \textbf{T}$$

$$e. B \perp\!\!\!\perp D \mid C \Leftrightarrow P(B, D \mid C) = \frac{P(B, D, C)}{P(C)} = \frac{P(D \mid C) \overbrace{P(B \mid C)}^{P(B, C)}}{P(C)}$$



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$$= \frac{P(D \mid C) \overbrace{P(B, C)}^{P(B \mid C)}}{P(C)}$$

$$= P(D \mid C) P(B \mid C)$$

\Rightarrow Factorizes

$\Rightarrow T$

2.1



GENERATIVE MODEL

$$\varphi_k \sim \text{Dirichlet}_N(\beta) \quad \text{for } k = 1 \dots T$$

$$\theta_i \sim \text{Dirichlet}_T(\alpha) \quad \text{for } i = 1 \dots D$$

for each word j , for $j = 1 \dots W$,

$$t_{ij} \mid \theta_i \sim \text{Categorical}(\theta_i)$$

$$w_{ij} \mid \varphi_{t_{ij}} \sim \text{Categorical}(\varphi_{t_{ij}})$$