

Inspector Clusseau

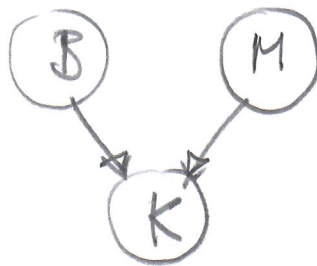
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$$\left. \begin{array}{l} p(b) = 0.6 \\ p(m) = 0.2 \end{array} \right\} \text{ priors, independent!}$$

Table of conditionals: (et complements)

$$\begin{aligned} p(k|\bar{b}, \bar{m}) &= 0.3 & \rightarrow p(\bar{k}|\bar{b}, \bar{m}) &= 1 - p(k|\bar{b}, \bar{m}) \\ p(k|\bar{b}, m) &= 0.2 & &= 0.7 \\ p(k|b, \bar{m}) &= 0.6 \\ p(k|b, m) &= 0.1 \end{aligned}$$

DAG: Directed Acyclic Graph



"is murder"



$$\begin{aligned} \text{dom}(M) &= \{m, \bar{m}\} \\ \text{dom}(B) &= \{b, \bar{b}\} \\ \text{dom}(K) &= \{k, \bar{k}\} \end{aligned}$$

⇒ joint probability distribution

$$\begin{aligned} p(K, B, M) &= p(K \wedge B \wedge M) \\ &= p(K|B, M) \cdot \underbrace{p(B) \cdot p(M)}_{\text{independent!}} \end{aligned}$$

⇒ 3D probability table

$$\sim 8 = 2^3 \text{ entries}$$

We look for:

$$p(\mathcal{Z}=b \mid k=k) =$$

$$p(b \mid k) = ?$$

no $M \rightarrow$ need to marginalize m out (average out...)

marginalize $m \in \text{dom}(M)$

$$p(b \mid k) = \sum_m p(b, M \mid k)$$

$$= \sum_m \frac{p(b, M, k)}{p(k)} \quad (\text{Bayes, Product Rule})$$

$$= \frac{\sum_m p(k \mid b, M) \cdot p(b) \cdot p(M)}{\sum_m \sum_b p(k \mid b, M) \cdot p(b) \cdot p(M)}$$

$p(k)$

$$= \frac{p(b) \cdot \sum_m p(k \mid b, M) \cdot p(M)}{\sum_{b \in \text{dom}(B)} p(b) \cdot \sum_{m \in \text{dom}(M)} p(M) \cdot p(k \mid b, M)}$$

Alternatively:

$$p(b \mid k) = \sum_m p(b, M \mid k)$$

$$= \sum_m \frac{p(k \mid b, M) p(b) p(M)}{p(k)}$$

$$p(k) = \sum_{b \in B} \sum_{m \in M} p(k, b, m) \quad \} \text{ "evidence" }$$