

Oplossingen Bayesian Networks

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6 november 2025

Problem 1: Host's Console Crash

Let's define the binary variables based on the problem description. Note the interpretation of P :

- $P = 1$: Power is OK, $P = 0$: Power problem.
- $G = 1$: Game is buggy, $G = 0$: Game is not buggy.
- $M = 1$: Music app playing, $M = 0$: Music app not playing.
- $C = 1$: Console crash, $C = 0$: Console not crashed.

The given probabilities are:

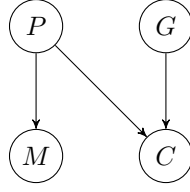
- $P(P = 1) = 0.97 \implies P(P = 0) = 0.03$
- $P(G = 1) = 0.35 \implies P(G = 0) = 0.65$
- $P(M = 1|P = 1) = 0.92 \implies P(M = 0|P = 1) = 0.08$
- $P(M = 1|P = 0) = 0.12 \implies P(M = 0|P = 0) = 0.88$
- $P(C = 1|P = 0, G = 1) = 0.98$
- $P(C = 1|P = 1, G = 1) = 0.45$
- $P(C = 1|P = 0, G = 0) = 0.88$
- $P(C = 1|P = 1, G = 0) = 0.08$

a) Bayesian Network and Factorization

The story implies:

- "P and G are a priori independent" $\implies P$ and G are roots.
- "Music... if power is fine" $\implies M$ depends on P .
- "Crashes... due to... power problem (P) or a buggy game (G)" $\implies C$ depends on both P and G .

The Bayesian Network diagram is:



The factorization of the joint probability distribution is:

$$P(P, G, M, C) = P(P) \cdot P(G) \cdot P(M|P) \cdot P(C|P, G)$$

b) Storage

- **Full Joint:** A full joint distribution over 4 binary variables has $2^4 = 16$ possible entries. To specify the distribution, we need $2^4 - 1 = 15$ independent parameters (the last one is constrained by the sum-to-one rule).
- **Factorization:** We count the independent parameters required for each Conditional Probability Table (CPT) in our factorization:
 - $P(P)$: 1 parameter ($P(P = 1)$).
 - $P(G)$: 1 parameter ($P(G = 1)$).
 - $P(M|P)$: 2 parameters ($P(M = 1|P = 1)$ and $P(M = 1|P = 0)$).
 - $P(C|P, G)$: 4 parameters (one for each of the $2^2 = 4$ combinations of parent states).

The total number of parameters is $1 + 1 + 2 + 4 = 8$.

c) Inference

(i) $P(G = 1|C = 1)$

We use Bayes' rule: $P(G = 1|C = 1) = \frac{P(C=1|G=1)P(G=1)}{P(C=1)}$.

First, we find $P(C = 1|G = 1)$ by marginalizing over P :

$$\begin{aligned} P(C = 1|G = 1) &= \sum_p P(C = 1|P = p, G = 1)P(P = p) \\ &= P(C = 1|1, 1)P(P = 1) + P(C = 1|0, 1)P(P = 0) \\ &= (0.45 \cdot 0.97) + (0.98 \cdot 0.03) = 0.4365 + 0.0294 = 0.4659 \end{aligned}$$

Next, we find the marginal $P(C = 1)$ using the law of total probability:

$$\begin{aligned} P(C = 1) &= \sum_g P(C = 1|G = g)P(G = g) \\ &= P(C = 1|G = 1)P(G = 1) + P(C = 1|G = 0)P(G = 0) \end{aligned}$$

We need $P(C = 1|G = 0) = \sum_p P(C = 1|p, 0)P(p)$:

$$\begin{aligned} P(C = 1|G = 0) &= P(C = 1|1, 0)P(P = 1) + P(C = 1|0, 0)P(P = 0) \\ &= (0.08 \cdot 0.97) + (0.88 \cdot 0.03) = 0.0776 + 0.0264 = 0.104 \end{aligned}$$

Now, plug this back into the equation for $P(C = 1)$:

$$P(C = 1) = (0.4659 \cdot 0.35) + (0.104 \cdot 0.65) = 0.163065 + 0.0676 = 0.230665$$

Finally, we compute the desired probability:

$$P(G = 1|C = 1) = \frac{P(C = 1|G = 1)P(G = 1)}{P(C = 1)} = \frac{0.4659 \cdot 0.35}{0.230665} = \frac{0.163065}{0.230665} \approx 0.7069$$

(ii) $P(G = 1|C = 1, M = 1)$

By definition of conditional probability: $P(G = 1|C = 1, M = 1) = \frac{P(G=1, C=1, M=1)}{P(C=1, M=1)}$.

Numerator: $P(G = 1, C = 1, M = 1) = \sum_p P(P = p, G = 1, M = 1, C = 1)$

$$= \sum_p P(G = 1)P(p)P(M = 1|p)P(C = 1|p, G = 1)$$

$$= P(G = 1) \cdot [P(P = 1)P(M = 1|1)P(C = 1|1, 1) + P(P = 0)P(M = 1|0)P(C = 1|0, 1)]$$

$$= 0.35 \cdot [(0.97 \cdot 0.92 \cdot 0.45) + (0.03 \cdot 0.12 \cdot 0.98)]$$

$$= 0.35 \cdot [0.40158 + 0.003528] = 0.35 \cdot 0.405108 = 0.1417878$$

Denominator: $P(C = 1, M = 1) = \sum_g P(G = g, C = 1, M = 1)$

$$= P(G = 1, C = 1, M = 1) + P(G = 0, C = 1, M = 1)$$

We just calculated the $G = 1$ term. Now for the $G = 0$ term:

$$P(G = 0, C = 1, M = 1) = P(G = 0) \cdot [P(P = 1)P(M = 1|1)P(C = 1|1, 0) + P(P = 0)P(M = 1|0)P(C = 1|0, 0)]$$

$$= 0.65 \cdot [(0.97 \cdot 0.92 \cdot 0.08) + (0.03 \cdot 0.12 \cdot 0.88)]$$

$$= 0.65 \cdot [0.071392 + 0.003168] = 0.65 \cdot 0.07456 = 0.048464$$

So, $P(C = 1, M = 1) = 0.1417878 + 0.048464 = 0.1902518$.

Finally:

$$P(G = 1|C = 1, M = 1) = \frac{0.1417878}{0.1902518} \approx 0.74526$$

d) Independence

Based on the network structure and d-separation rules:

- **Independent pairs ($X \perp Y$):**

- (G, P) : **Independent.** They are two roots with no path between them.
- (G, M) : **Independent.** The only path is $G \rightarrow C \leftarrow P \rightarrow M$, which is blocked by the collider C .

- **Conditionally independent given exactly one variable ($X \perp Y|Z$):**

- $(G, M)|P$: **Independent.** The path $G \rightarrow C \leftarrow P \rightarrow M$ is blocked by C (collider) and also by the given P (which blocks the $P \rightarrow M$ part).
- $(C, M)|P$: **Independent.** The path $C \leftarrow P \rightarrow M$ is blocked by P . The path $C \leftarrow G$ is not relevant.
- $(P, G)|M$: **Independent.** The path $P \rightarrow C \leftarrow G$ is blocked by the collider C . (M is not C or a descendant).

Problem 2: Party Noise Network

a) True/False using d-separation

- (i) $h \perp s$: **True.** The path $h \rightarrow w \rightarrow e \leftarrow l \leftarrow s$ is blocked by the collider e . The path $h \rightarrow w \rightarrow e \rightarrow d \leftarrow b \leftarrow s$ is blocked by the collider d .
- (ii) $h \perp l$: **True.** The path $h \rightarrow w \rightarrow e \leftarrow l$ is blocked by the collider e .
- (iii) $h \perp x$: **False.** The path $h \rightarrow w \rightarrow e \rightarrow x$ is an active path (a chain).
- (iv) $w \perp s$: **True.** The path $w \rightarrow e \leftarrow l \leftarrow s$ is blocked by collider e . The path $w \rightarrow e \rightarrow d \leftarrow b \leftarrow s$ is blocked by collider d .
- (v) $x \perp b$: **True.** The path $x \leftarrow e \rightarrow d \leftarrow b$ is blocked by collider d . The path $x \leftarrow e \leftarrow l \leftarrow s \rightarrow b$ is blocked at e (collider) and s (common cause, but path from e is blocked).

b) Compute

We need to compute $P(d = 1)$, $P(d = 1|s = 1)$, and $P(d = 1|s = 0)$. We will use the structure: $P(d = 1|s) = \sum_{e,b} P(d = 1|e, b)P(e, b|s)$. From the graph, $e \perp b|s$. (The path $e \leftarrow l \leftarrow s \rightarrow b$ is blocked by s . The other path $e \leftarrow w \leftarrow h$ is independent of s and b). Therefore, $P(d = 1|s) = \sum_{e,b} P(d = 1|e, b)P(e|s)P(b|s)$.

First, let's find the components $P(e|s)$ and $P(b|s)$. $P(b|s)$ is given. For $P(e|s)$: $e = w \vee l$, so $e = 0 \iff w = 0 \text{ and } l = 0$. $P(e = 0|s) = P(w = 0, l = 0|s) = P(w = 0|s)P(l = 0|s)$. w is d-separated from s , so $P(w = 0|s) = P(w = 0)$. $P(w = 1) = P(w|h = 1)P(h = 1) + P(w|h = 0)P(h = 0)$. $P(w = 1) = (0.5 \cdot 0.2) + (0.1 \cdot 0.8) = 0.1 + 0.08 = 0.18$. $P(w = 0) = 1 - 0.18 = 0.82$. So, $P(e = 0|s) = 0.82 \cdot P(l = 0|s)$.

Calculations for $s = 1$:

- $P(l = 0|s = 1) = 1 - P(l = 1|s = 1) = 1 - 0.3 = 0.7$.
- $P(e = 0|s = 1) = P(w = 0)P(l = 0|s = 1) = 0.82 \cdot 0.7 = 0.574$.
- $P(e = 1|s = 1) = 1 - 0.574 = 0.426$.
- $P(b = 1|s = 1) = 0.6$, $P(b = 0|s = 1) = 0.4$.

Calculations for $s = 0$:

- $P(l = 0|s = 0) = 1 - P(l = 1|s = 0) = 1 - 0.05 = 0.95$.
- $P(e = 0|s = 0) = P(w = 0)P(l = 0|s = 0) = 0.82 \cdot 0.95 = 0.779$.
- $P(e = 1|s = 0) = 1 - 0.779 = 0.221$.
- $P(b = 1|s = 0) = 0.2$, $P(b = 0|s = 0) = 0.8$.

Now we can compute the probabilities for $d = 1$. Given $P(d = 1|e, b)$: $P(d|1, 1) = 0.9$, $P(d|1, 0) = 0.7$, $P(d|0, 1) = 0.8$, $P(d|0, 0) = 0.1$.

(ii) $P(d = 1|s = 1)$

$$\begin{aligned} P(d = 1|s = 1) &= \sum_{e,b} P(d|e, b)P(e|s = 1)P(b|s = 1) \\ &= P(d|1, 1)P(e = 1|1)P(b = 1|1) + P(d|1, 0)P(e = 1|1)P(b = 0|1) + \\ &\quad P(d|0, 1)P(e = 0|1)P(b = 1|1) + P(d|0, 0)P(e = 0|1)P(b = 0|1) \\ &= (0.9 \cdot 0.426 \cdot 0.6) + (0.7 \cdot 0.426 \cdot 0.4) + (0.8 \cdot 0.574 \cdot 0.6) + (0.1 \cdot 0.574 \cdot 0.4) \\ &= 0.23004 + 0.11928 + 0.27552 + 0.02296 = \mathbf{0.6478} \end{aligned}$$

(iii) $P(d = 1|s = 0)$

$$\begin{aligned} P(d = 1|s = 0) &= \sum_{e,b} P(d|e, b)P(e|s = 0)P(b|s = 0) \\ &= P(d|1, 1)P(e = 1|0)P(b = 1|0) + P(d|1, 0)P(e = 1|0)P(b = 0|0) + \\ &\quad P(d|0, 1)P(e = 0|0)P(b = 1|0) + P(d|0, 0)P(e = 0|0)P(b = 0|0) \\ &= (0.9 \cdot 0.221 \cdot 0.2) + (0.7 \cdot 0.221 \cdot 0.8) + (0.8 \cdot 0.779 \cdot 0.2) + (0.1 \cdot 0.779 \cdot 0.8) \\ &= 0.03978 + 0.12376 + 0.12464 + 0.06232 = \mathbf{0.3505} \end{aligned}$$

(i) $P(d = 1)$

Using the law of total probability with $P(s = 1) = 0.5$ and $P(s = 0) = 0.5$:

$$\begin{aligned} P(d = 1) &= P(d = 1|s = 1)P(s = 1) + P(d = 1|s = 0)P(s = 0) \\ &= (0.6478 \cdot 0.5) + (0.3505 \cdot 0.5) \\ &= 0.5 \cdot (0.6478 + 0.3505) = 0.5 \cdot 0.9983 = \mathbf{0.49915} \end{aligned}$$

Problem 3: D-Separation Quick Test

1. $X \rightarrow Z \rightarrow Y$ (Chain)

- $X \perp Y$? **No.** The path $X \rightarrow Z \rightarrow Y$ is active.
- $X \perp Y|Z$? **Yes.** The path $X \rightarrow Z \rightarrow Y$ is blocked by conditioning on Z .

2. $X \leftarrow Z \rightarrow Y$ (Common cause)

- $X \perp Y$? **No.** The path $X \leftarrow Z \rightarrow Y$ is active.
- $X \perp Y|Z$? **Yes.** The path $X \leftarrow Z \rightarrow Y$ is blocked by conditioning on Z .

3. $X \rightarrow Z \leftarrow Y$ (Collider / v-structure)

- $X \perp Y$? **Yes.** The path $X \rightarrow Z \leftarrow Y$ is blocked by the collider Z (which is not conditioned on).
- $X \perp Y|Z$? **No.** The path is opened by conditioning on the collider Z .

4. $X \leftarrow Z \leftarrow Y$ (Chain)

- $X \perp Y$? **No.** The path $Y \rightarrow Z \rightarrow X$ is active.
- $X \perp Y|Z$? **Yes.** The path $Y \rightarrow Z \rightarrow X$ is blocked by conditioning on Z .