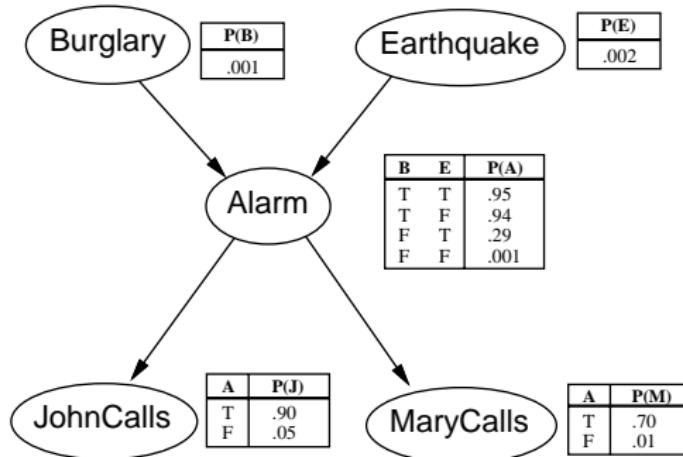


Variable Elimination Example

A typical belief network with conditional probabilities is given in the following figure:



The letters B, E, A, J , and M stand for *Burglary*, *Earthquake*, *Alarm*, *JohnCalls*, and *MaryCalls*, respectively. All variables (nodes) are Boolean, so the probability of, say, $\neg P(A)$ in any row of its table is $1 - P(A)$. Use *variable elimination* method to compute probability $P(B|j, m)$.

| a | $P(m a) = f_M(a)$ |
|---|-------------------|
|---|-------------------|

| | |
|---|------|
| t | 0.70 |
|---|------|

| | |
|---|------|
| f | 0.01 |
|---|------|

| a | $P(j a) = f_J(a)$ |
|---|-------------------|
|---|-------------------|

| | |
|---|------|
| t | 0.90 |
|---|------|

| | |
|---|------|
| f | 0.05 |
|---|------|

| a | b | e | $P(a b, e) = f_A(a, b, e)$ |
|---|---|---|----------------------------|
|---|---|---|----------------------------|

| | | | |
|---|---|---|------|
| t | t | t | 0.95 |
|---|---|---|------|

| | | | |
|---|---|---|------|
| t | t | f | 0.94 |
|---|---|---|------|

| | | | |
|---|---|---|------|
| t | f | t | 0.94 |
|---|---|---|------|

| | | | |
|---|---|---|------|
| f | t | t | 0.05 |
|---|---|---|------|

| | | | |
|---|---|---|------|
| f | t | f | 0.06 |
|---|---|---|------|

| | | | |
|---|---|---|------|
| f | f | t | 0.71 |
|---|---|---|------|

| | | | |
|---|---|---|-------|
| f | f | f | 0.999 |
|---|---|---|-------|

| b | e | $f_{\bar{A}JM}(b, e)$ |
|---|---|---|
| t | t | $0.95*0.70*0.90 + 0.05*0.01*0.05 = 0.598525$ |
| t | f | $0.94*0.70*0.90 + 0.06*0.01*0.05 = 0.59223$ |
| f | t | $0.29*0.70*0.90 + 0.71*0.01*0.05 = 0.183055$ |
| f | f | $0.001*0.70*0.90 + 0.999*0.01*0.05 = 0.0011295$ |

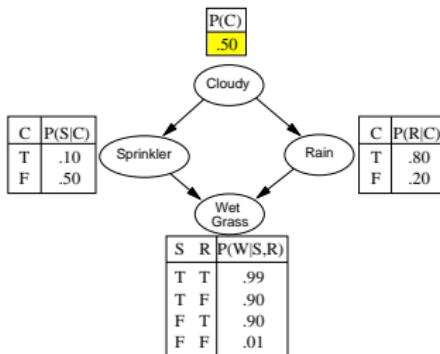
| e | $P(e) = f_E(e)$ | b | $f_{E\bar{A}JM}(e)$ |
|---|-----------------|---|--|
| t | 0.002 | t | $0.002*0.598525 + 0.998*0.59223 = 0.59224259$ |
| f | 0.998 | f | $0.002*0.183055 + 0.998*0.0011295 = 0.0014934$ |

| b | $f_B(b)$ | b | $f_B(b) \times f_{E\bar{A}JM}(e)$ |
|---|----------|---|------------------------------------|
| t | 0.001 | t | $0.001*0.59224259 = 0.00059224259$ |
| f | 0.999 | f | $0.999*0.0014934 = 0.0014919066$ |

$$\mathbf{P}(\mathbf{B}|j, m) = \alpha < 0.00059224259, 0.0014919066 > \approx < 0.2842, 0.7158 >$$

Problem #24

A typical belief network with conditional probabilities is given in the following figure:



The letters C, R, S and W stand for *Cloudy*, *Rain*, *Sprinkler*, and *Wet Grass*, respectively. All variables (nodes) are Boolean, so the probability of, say, $\neg A$ in any row of its table is $1 - P(A)$.

- Assuming that the nodes are introduced in the following order *Wet Grass*, *Sprinkler*, *Rain* and *Cloudy* construct a corresponding belief network. Show which probabilities need to be specified.
- Compute probabilities $P(W)$ and $P(S|W)$.

Problem #24: Solution a)

Order: W, S, R, C . Probabilities $P(W), P(S|W), P(R|S, W), P(C|R, S, W)$

- $P(W)$

Problem #24: Solution a)

Order: W, S, R, C . Probabilities $P(W), P(S|W), P(R|S, W), P(C|R, S, W)$

- $P(W)$
- $P(S|W) = P(S)?$

Problem #24: Solution a)

Order: W, S, R, C . Probabilities $P(W), P(S|W), P(R|S, W), P(C|R, S, W)$

- $P(W)$
- $P(S|W) = P(S)?$
- $P(S|W) = P(S)?$ No!

Problem #24: Solution a)

Order: W, S, R, C . Probabilities $P(W), P(S|W), P(R|S, W), P(C|R, S, W)$

- $P(W)$
- $P(S|W) = P(S)?$
- $P(S|W) = P(S)?$ No!
- $P(R|S, W) = P(R|S)?$

Problem #24: Solution a)

Order: W, S, R, C . Probabilities $P(W), P(S|W), P(R|S, W), P(C|R, S, W)$

- $P(W)$
- $P(S|W) = P(S)?$
- $P(S|W) = P(S)?$ No!
- $P(R|S, W) = P(R|S)?$
- $P(R|S, W) = P(R|S)?$ No!

Problem #24: Solution a)

Order: W, S, R, C . Probabilities $P(W), P(S|W), P(R|S, W), P(C|R, S, W)$

- $P(W)$
- $P(S|W) = P(S)?$
- $P(S|W) = P(S)? \text{ No!}$
- $P(R|S, W) = P(R|S)?$
- $P(R|S, W) = P(R|S)? \text{ No!}$
- $P(C|R, S, W) = P(C|R, S)? \quad P(C|R, S, W) = P(C|R)?$
- $P(C|R, S, W) = P(C|R, S)? \text{ Yes!} \quad P(C|R, S, W) = P(C|R)? \text{ No!}$

Problem #24: Solution b)

$$P(W) = \sum_c \sum_s \sum_r P(c)P(s|c)P(r|c)P(W|s, r)$$

$$P(W) = \sum_c P(c) \sum_s P(s|c) \sum_r P(r|c)P(W|s, r)$$

$$\begin{aligned} P(w) &= P(c)\{P(s|c)[P(r|c)P(w|s, r) + P(\neg r|c)P(w|s \neg r)] + \\ &\quad P(\neg s|c)[P(r|c)P(w|\neg s, r) + P(\neg r|c)P(w|\neg s \neg r)]\} + \\ &\quad P(\neg c)\{P(s|\neg c)[P(r|\neg c)P(w|s, r) + P(\neg r|\neg c)P(w|s \neg r)] + \\ &\quad P(\neg s|\neg c)[P(r|\neg c)P(w|\neg s, r) + P(\neg r|\neg c)P(w|\neg s \neg r)]\} \end{aligned}$$

$$\begin{aligned} P(w) &= 0.5\{0.1[0.8 * 0.99 + 0.2 * 0.9] + 0.9[0.8 * 0.9 + 0.2 * 0.01]\} + \\ &\quad 0.5\{0.5[0.2 * 0.99 + 0.8 * 0.9] + 0.5[0.2 * 0.9 + 0.8 * 0.01]\} \end{aligned}$$

$$\begin{aligned} P(w) &= 0.5[0.1 * 1.512 + 0.9 * 0.722] + 0.5[0.5 * 0.918 + 0.5 * 0.188] \\ &\quad 0.5 * 0.801 + 0.5 * 0.553 = 0.677 \end{aligned}$$

$$P(\neg w) = 0.323$$

Problem #24: Solution b)

$$P(S|W) = P(S, W)/P(W) = \alpha \sum_c \sum_r P(c)P(S|c)P(r|c)P(W|S, r)$$

$$P(S|W) = \alpha \sum_c P(c)P(S|c) \sum_r P(r|c)P(W|S, r)$$

$$\begin{aligned} P(s|w) &= \alpha \{ P(c)P(s|c)[(P(r|c)P(w|s, r) + P(\neg r|c)P(w|s, \neg r)] + \\ &\quad P(\neg c)P(s|\neg c)[(P(r|\neg c)P(w|s, r) + P(\neg r|\neg c)P(w|s, \neg r)]\} \\ &= \alpha[0.05(0.8 * 0.99 + 0.18) + 0.25(0.2 * 0.99 + 0.72)] = 0.2781 \end{aligned}$$

$$P(\neg s|w) = \alpha[0.45(0.8 * 0.9 + 0.2 * 0.01) + 0.25(0.2 * 0.9 + 0.8 * 0.01)]$$

$$P(s|\neg w) = ?$$

$$P(\neg s|\neg w) = ?$$

$$P(s|w) + P(\neg s|w) = 1, \quad P(s|\neg w) + P(\neg s|\neg w) = 1,$$