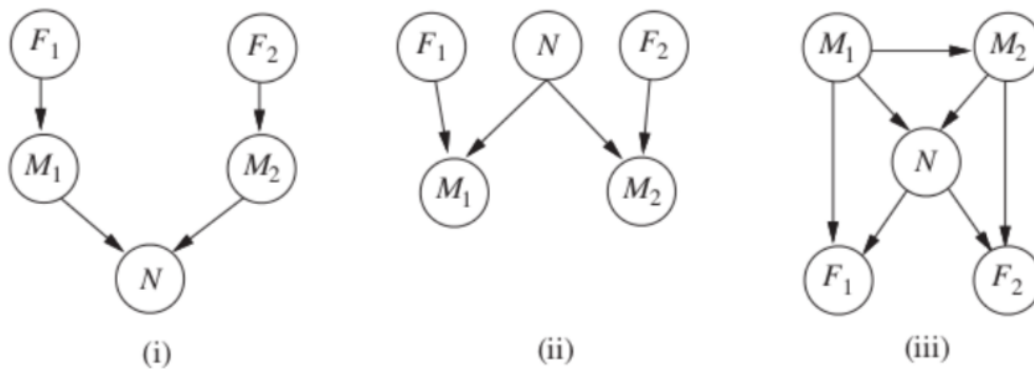


Name:

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1. (14 points) Two astronomers are making measurements M_1 and M_2 of the number of stars N in a small region of sky using telescopes. The measurements are noisy, so there is a probability of e that the measurements are off by ± 1 . Besides this, the telescope might be out of focus with probability f ($F_i = \text{true}$ means the i -th astronomer's telescope is out of focus, $i = 1, 2$). If it is out of focus, then the measurement will undercount N by at least 3 (or if $N < 3$, then M_i will just be 0). Consider the following.



- a. (2 points) Which is the best network? Explain. (Hint: Accuracy and efficiency)

Best network: ii

Reason:

(i) suggests N (number of stars) is conditionally independent of F_i (telescope out of focus or not) given measurements M_i , which is incorrect from our description.
 (iii) indicates some more information not described (the arrow from M_1 to M_2)

- b. (9 points) Assuming using network (ii), and $N \in \{1, 2, 3\}$ and $M_1 \in \{0, 1, 2, 3, 4\}$, write out the conditional distribution for $P(M_1|N)$ in terms of e and f . (Fill the blanks and equation)

We compute $P(M_1|N)$ by summing over (marginalizing) F_1 (product rule and total probability).

$$P(M_1|N) = P(M_1|N, F_1)P(F_1|N) + P(M_1|N, \neg F_1)P(\neg F_1|N) = P(M_1|N, F_1)P(F_1) + P(M_1|N, \neg F_1)P(\neg F_1)$$

$P(M_1 N)$	$N = 1$	$N = 2$	$N = 3$
$M_1 = 0$	$f+e(1-f)$	f	f
$M_1 = 1$	$(1-2e)(1-f)$	$e(1-f)$	0
$M_1 = 2$	$e(1-f)$	$(1-2e)(1-f)$	$e(1-f)$
$M_1 = 3$	0	$e(1-f)$	$(1-2e)(1-f)$
$M_1 = 4$	0	0	$e(1-f)$

$$P(M_1|N) = P(M_1, N)/P(N)$$

$$P(M_1, N) = P(M_1, N, F_1) + P(M_1, N, \neg F_1)$$

$$= P(M_1|N, F_1)P(F_1|N)P(N) + P(M_1|N, \neg F_1)P(\neg F_1|N)P(N)$$

$$P(M_1|N) = (P(M_1|N, F_1)P(F_1|N)P(N) + P(M_1|N, \neg F_1)P(\neg F_1|N)P(N))/P(N)$$

$$= P(M_1|N, F_1)P(F_1|N) + P(M_1|N, \neg F_1)P(\neg F_1|N)$$

$$= P(M_1|N, F_1)P(F_1) + P(M_1|N, \neg F_1)P(\neg F_1) \quad \text{F1 and N are independent}$$

$$P(M_1, N, F_1) = P(M_1|N, F_1)P(N, F_1) = P(M_1|N, F_1)P(F_1|N)P(N)$$

$$= P(M_1|N, F_1)P(N|F_1)P(F_1)$$

$$M_1=2, N=2$$

$$P(M_1|N) = P(M_1|N, F_1)P(F_1) + P(M_1|N, \neg F_1)P(\neg F_1)$$

$$P(F_1) = f$$

$$P(M_1|N, F_1) = 0$$

$$P(\neg F_1) = 1-f$$

$$P(M_1|N, \neg F_1) = 1-2e$$

$$P(M_1=N=1) = e \quad P(M_1=N=-1) = e \quad P(M_1=N) = 1-2e$$

$$\text{If } F_1=\text{true}, N=3, M_1=0$$

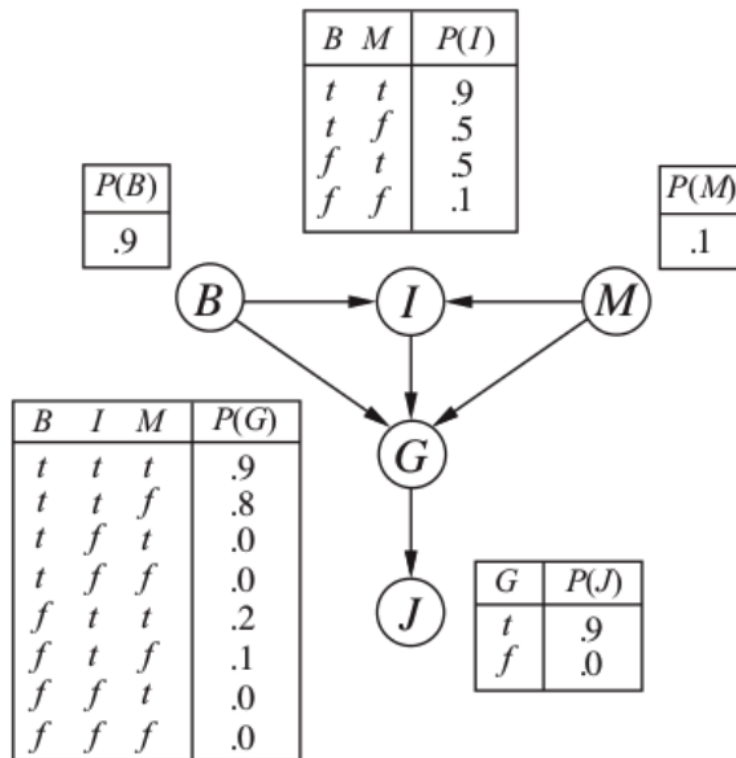
$$P(M_1|N=1, F_1=\text{true}) = 1$$

- c. (3 points) Suppose $M_1 = 1$ and $M_2 = 3$, and assume no prior constraint on N . What are the possible numbers that N can be? Why? (Hint: fix $M_1 = 1$, $M_2 = 3$, try each possibility of F_1 , F_2 and get all possible values of N .)

start from $N = 0, 1, \dots$, and for each value, see whether $M1 = 1$ and $M2 = 3$ are possible with all possibilities of $F1, F2$

the solution should be $N = 2, 4$ or any number ≥ 6 .

2. (6 points) Consider this figure for the next problem.



- a. (3 points) Which of the following (if any) are asserted by the network structure (ignoring the conditional probability tables)? (Hint: Write "Yes" or "No" following each equation)

$P(B, I, M) = P(B)P(I)P(M)$; No

$P(J, G) = P(J|G, I)$; No

$P(M|G, B, I) = P(M|G, B, I, J)$; Yes

Markov blanket: a node is conditionally independent of all other nodes in the network, given its parents, children, and children's parents

- b. (3 points) Calculate the value of $P(b, i, m, \neg g, j)$. (Fill the blanks and compute the result)

$$P(b, i, m, \neg g, j) = P(\quad)P(m)P(i|b, m)P(\quad)P(j|\neg g) =$$

$$P(b, i, m, \neg g, j) = P(b)P(m)P(i|b, m)P(\neg g|b, i, m)P(j|\neg g) = 0.9 \times 0.1 \times 0.9 \times (1 - 0.9) \times 0.0 = 0$$

Global semantics: $P(X_2, X_1)$ joint probability

Marginalization: $P(X_1) = \text{sum over } X_2, P(X_1, X_2)$

Product: $P(X_2, X_1) = P(X_2|X_1)P(X_1)$ two events/semantics

$P(X_2|X_1) = P(X_2, X_1)/P(X_1)$

$P(x_1, \dots, x_n) = P(x_n | x_{n-1}, \dots, x_1)P(x_{n-1}, \dots, x_1)$ n events/semantics

Goal: $P(x_1, \dots, x_n) = P(x_n | x_{n-1}, \dots, x_1)P(x_{n-1} | x_{n-2}, \dots, x_1) \cdots P(x_2 | x_1)P(x_1)$

$P(X_2, X_1) = P(X_2|X_1)P(X_1)$ True

$P(X_3, X_2, X_1) = P(X_3, X_2 | X_1)P(X_1)$

$P(X_3, X_2, X_1) = P(X_3 | X_2, X_1)P(X_2, X_1)$ True

$P(X_3, X_2, X_1) = P(X_3 | X_2, X_1)P(X_2|X_1)P(X_1)$

$P(X_4, X_3, X_2, X_1) = P(X_4 | X_3, X_2, X_1)P(X_3 | X_2, X_1)P(X_2|X_1)P(X_1)$

...

$P(x_n, \dots, x_1) = P(x_n | x_{n-1}, \dots, x_1)P(x_{n-1} | x_{n-2}, \dots, x_1)P(x_{n-2} | x_{n-3}, \dots, x_1)P(x_{n-3}, \dots, x_1)$

$= P(x_n | x_{n-1}, \dots, x_1)P(x_{n-1} | x_{n-2}, \dots, x_1) \cdots P(x_2 | x_1)P(x_1)$

Independence:

If A, B are independent, $P(A, B) = P(A)P(B)$; $P(A|B) = P(A)$

Generally, $P(A, B) = (\# \text{ of } (A=\text{True}) \& (B=\text{True})) / ((\# \text{ of } (A=\text{True}) \& (B=\text{True})) + (\# \text{ of } (A=\text{True}) \& (B=\text{False})) + (\# \text{ of } (A=\text{False}) \& (B=\text{True})) + (\# \text{ of } (A=\text{False}) \& (B=\text{False})))$

If independent: $P(A, B) = (\# \text{ of } (A=\text{True}) / \# \text{ all experiments } (A=\text{True} \text{ or } A=\text{False})) * (\# \text{ of } (B=\text{True}) / \# \text{ all experiments } (B=\text{True} \text{ or } B=\text{False}))$

$P(A|B) = (\# \text{ of } (A=\text{True}) \& (B=\text{True})) / (\# (B=\text{True}))$

1b

$P(M1, N) = P(M1, N, F1) + P(M1, N, !F1)$

#T times experiments

$P(M1, N) = \# (M1=m \text{ and } N=n) / \#T$

$= (\# (M1=m \text{ and } N=n \text{ and } F1=\text{true}) + \# (M1=m \text{ and } N=n \text{ and } F1=\text{false})) / \#T$

$= \# (M1=m \text{ and } N=n \text{ and } F1=\text{true}) / \#T + \# (M1=m \text{ and } N=n \text{ and } F1=\text{false}) / \#T$

$= P(M1, N, F1) + P(M1, N, !F1)$

$= P(M1|N, F1)P(F1|N)P(N) + P(M1|N, !F1)P(!F1|N)P(N)$

$$P(M1 | N) = P(M1, N) / P(N)$$

$$F1 \text{ and } N \text{ independent, } P(F1|N) = P(F1)$$

#T

$$P(A) = \#(A \text{ is true}) / \#T$$

$$P(M1-N==1) = e$$

$$P(M1-N== -1) = e$$

1c

$$P(N|M1=1, M2=3)$$

$$\text{Valid } N=n \text{ means } P(N=n|M1=1, M2=3) > 0$$