

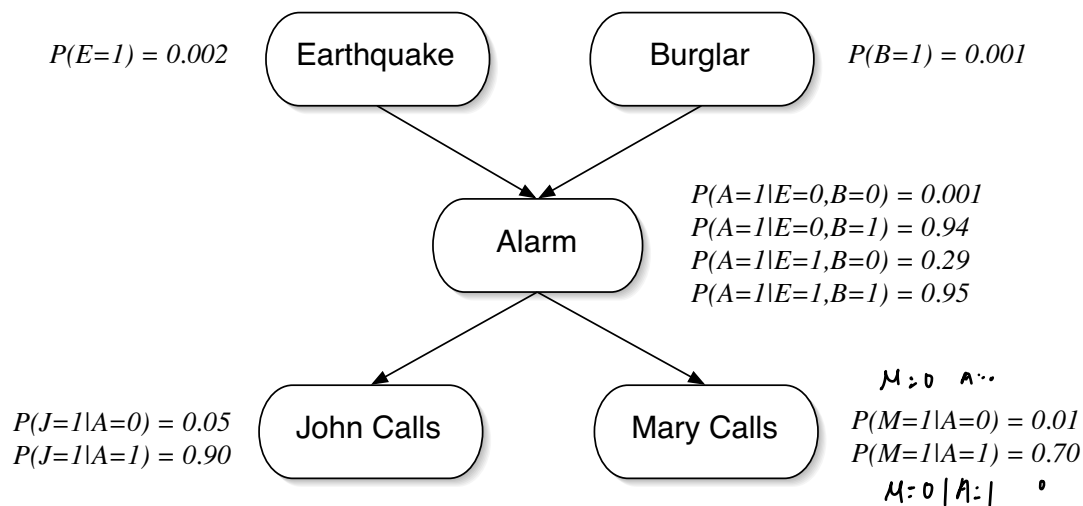
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**CSE 150A / 250A - Homework 2 (51 pts total)****Out:** Tue Oct 8**Due:** Mon Oct 14 (by 11:59 PM, Pacific Time, via gradescope)**Grace period:** 24 hours

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✓ **2.1 Probabilistic inference (12 pts — 2 pts each)**

Recall the alarm belief network described in class. The directed acyclic graph (DAG) and conditional probability tables (CPTs) are shown below:



Compute numeric values for the following probabilities, exploiting relations of marginal and conditional independence as much as possible to simplify your calculations. You may re-use numerical results from lecture, but otherwise *show your work*. Be careful not to drop significant digits in your answer.

- |                       |                       |                       |
|-----------------------|-----------------------|-----------------------|
| (a) $P(E=1 A=1)$      | (c) $P(A=1 M=1)$      | (e) $P(A=1 M=0)$      |
| (b) $P(E=1 A=1, B=0)$ | (d) $P(A=1 M=1, J=0)$ | (f) $P(A=1 M=0, B=1)$ |

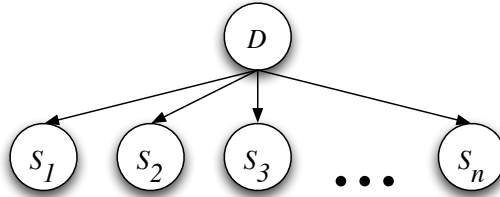
Consider your results in (b) versus (a), (d) versus (c), and (f) versus (e). Do they seem consistent with commonsense patterns of reasoning?

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## 2.2 Probabilistic reasoning (4 pts)

A patient is known to have contracted a rare disease which comes in two forms, represented by the values of a binary random variable  $D \in \{0, 1\}$ . Symptoms of the disease are represented by the binary random variables  $S_k \in \{0, 1\}$ , and knowledge of the disease is summarized by the belief network:



The conditional probability tables (CPTs) for this belief network are as follows. In the absence of evidence, both forms of the disease are equally likely, with prior probabilities:

$$P(D=0) = P(D=1) = \frac{1}{2}.$$

In one form of the disease ( $D=0$ ), the first symptom occurs with probability one,

$$P(S_1=1|D=0) = 1,$$

while the  $k^{\text{th}}$  symptom (with  $k \geq 2$ ) occurs with probability

$$P(S_k=1|D=0) = \frac{f(k-1)}{f(k)},$$

where the function  $f(k)$  is defined by

$$f(k) = 2^k + (-1)^k.$$

By contrast, in the other form of the disease ( $D=1$ ), all the symptoms are uniformly likely to be observed, with

$$P(S_k=1|D=1) = \frac{1}{2}$$

for all  $k$ . Suppose that on the  $k^{\text{th}}$  day of the month, a test is done to determine whether the patient is exhibiting the  $k^{\text{th}}$  symptom, and that each such test returns a positive result. Thus, on the  $k^{\text{th}}$  day, the doctor observes the patient with symptoms  $\{S_1=1, S_2=1, \dots, S_k=1\}$ . Based on the cumulative evidence, the doctor makes a new diagnosis each day by computing the ratio:

$$r_k = \frac{P(D=0|S_1=1, S_2=1, \dots, S_k=1)}{P(D=1|S_1=1, S_2=1, \dots, S_k=1)}.$$

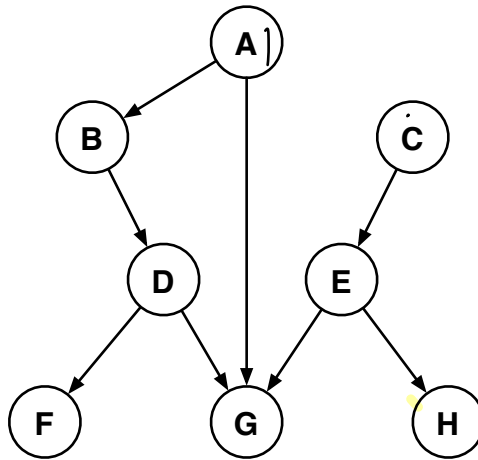
If this ratio is greater than 1, the doctor diagnoses the patient with the  $D=0$  form of the disease; otherwise, with the  $D=1$  form.

- Compute the ratio  $r_k$  as a function of  $k$ . How does the doctor's diagnosis depend on the day of the month? Show your work. (3 pts)
  - Does the diagnosis become more or less certain as more symptoms are observed? Explain. (1 pts)
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### 2.3 True or false (10 pts)

For the belief network shown below, indicate whether the following statements of marginal or conditional independence are **true (T)** or **false (F)**. Your answer will be **only** graded based on correctness. No justifications required.



\_\_\_\_\_

$$P(B|G, C) = P(B|G)$$

\_\_\_\_\_

$$P(F, G|D) = P(F|D) P(G|D)$$

\_\_\_\_\_

$$P(A, C) = P(A) P(C)$$

\_\_\_\_\_

$$P(D|B, F, G) = P(D|B, F, G, A)$$

\_\_\_\_\_

$$P(F, H) = P(F) P(H)$$

\_\_\_\_\_

$$P(D, E|F, H) = P(D|F) P(E|H)$$

\_\_\_\_\_

$$P(F, C|G) = P(F|G) P(C|G)$$

\_\_\_\_\_

$$P(D, E, G) = P(D) P(E) P(G|D, E)$$

\_\_\_\_\_

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

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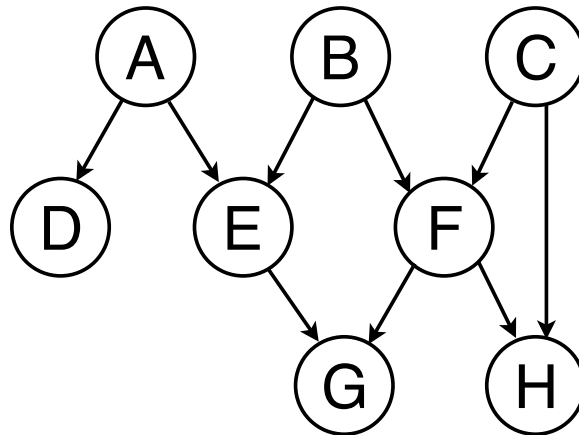
$$P(H|A, C) = P(H|A, C, G)$$

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## 2.4 More on Belief Networks (10 pts)

For the belief network shown below, indicate whether the following statements of marginal or conditional independence are **true (T)** or **false (F)**. Your answer will be **only** graded based on correctness. No justifications required.

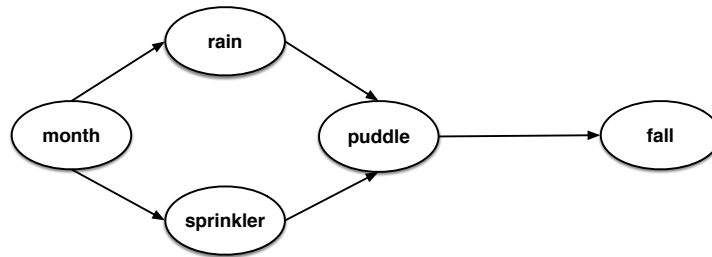
- (a) \_\_\_\_\_  $P(F|H) = P(F|C, H)$
- (b) \_\_\_\_\_  $P(E|A, B) = P(E|A, B, F)$
- (c) \_\_\_\_\_  $P(E, F|B, G) = P(E|B, G) P(F|B, G)$
- (d) \_\_\_\_\_  $P(F|B, C, G, H) = P(F|B, C, E, G, H)$
- ✓(e) \_\_\_\_\_  $P(A, B|D, E, F) = P(A, B|D, E, F, G, H)$
- (f) \_\_\_\_\_  $P(D, E, F) = P(D) P(E|D) P(F|E)$
- (g) \_\_\_\_\_  $P(A|F) = P(A)$
- ✓(h) \_\_\_\_\_  $P(E, F) = P(E) P(F)$
- (i) \_\_\_\_\_  $P(D|A) = P(D|A, E)$
- (j) \_\_\_\_\_  $P(B, C) = P(B) P(C)$



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## 2.5 Conditional independence (8 pts)

Consider the DAG shown below, describing the following domain. Given the month of the year, there is some probability of `rain`, and also some probability that the `sprinkler` is turned on. Either of these events leads to some probability that a `puddle` forms on the sidewalk, which in turn leads to some probability that someone has a `fall`.

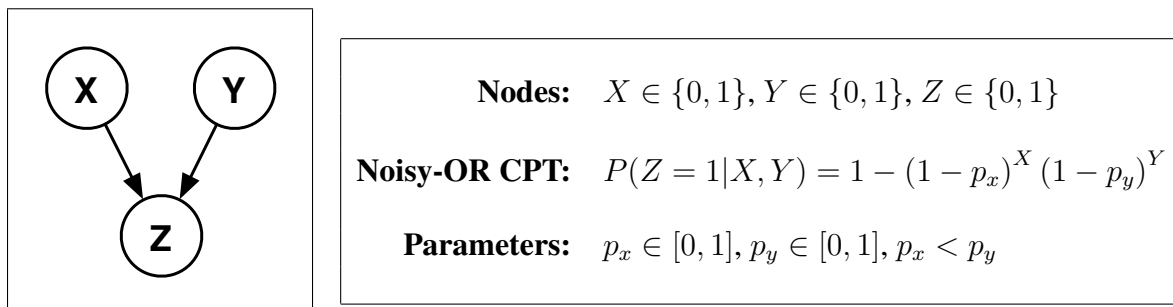


List all the conditional independence relations that must hold in any probability distribution represented by this DAG. More specifically, list all tuples  $\{X, Y, E\}$  such that  $P(X, Y|E) = P(X|E)P(Y|E)$ , where

$$\begin{aligned} X, Y &\in \{\text{month}, \text{rain}, \text{sprinkler}, \text{puddle}, \text{fall}\}, \\ E &\subseteq \{\text{month}, \text{rain}, \text{sprinkler}, \text{puddle}, \text{fall}\}, \\ X &\neq Y, \\ X, Y &\notin E. \end{aligned}$$

**Hint:** There are sixteen such tuples, not counting those that are equivalent up to exchange of  $X$  and  $Y$ . Do any of the tuples contain the case  $E = \emptyset$ ?

## 2.6 Noisy-OR (7 pts)



Suppose that the nodes in this network represent binary random variables and that the CPT for  $P(Z|X, Y)$  is parameterized by a noisy-OR model, as shown above. Suppose also that

$$0 < P(X=1) < 1,$$

$$0 < P(Y=1) < 1,$$

while the parameters of the noisy-OR model satisfy:

$$0 < p_x < p_y < 1.$$

Consider the following pairs of probabilities. In each case, indicate whether the probability on the left is equal (=), greater than (>), or less than (<) the probability on the right. The first one has been filled in for you as an example. (You should use your intuition for these problems; you are **not** required to show work.)

	$P(X=1)$	<div style="border: 1px solid black; padding: 2px 10px;">=</div>	$P(X=1)$
(a)	$P(Z=1 X=0, Y=0)$	<div style="border: 1px solid black; width: 40px; height: 30px; display: inline-block;"></div>	$P(Z=1 X=0, Y=1)$
(b)	$P(Z=1 X=1, Y=0)$	<div style="border: 1px solid black; width: 40px; height: 30px; display: inline-block;"></div>	$P(Z=1 X=0, Y=1)$
(c)	$P(Z=1 X=1, Y=0)$	<div style="border: 1px solid black; width: 40px; height: 30px; display: inline-block;"></div>	$P(Z=1 X=1, Y=1)$
(d)	$P(X=1)$	<div style="border: 1px solid black; width: 40px; height: 30px; display: inline-block;"></div>	$P(X=1 Z=1)$
(e)	$P(X=1)$	<div style="border: 1px solid black; width: 40px; height: 30px; display: inline-block;"></div>	$P(X=1 Y=1)$
(f)	$P(X=1 Z=1)$	<div style="border: 1px solid black; width: 40px; height: 30px; display: inline-block;"></div>	$P(X=1 Y=1, Z=1)$
(g)	$P(X=1) P(Y=1) P(Z=1)$	<div style="border: 1px solid black; width: 40px; height: 30px; display: inline-block;"></div>	$P(X=1, Y=1, Z=1)$

**Challenge (optional):** for each case, prove rigorously the correctness of your answer. You will receive partial credits for your efforts on proofs.